

STOCHASTIC SIX-VERTEX MODEL IN A HALF-QUADRANT AND HALF-LINE OPEN ASYMMETRIC SIMPLE EXCLUSION PROCESS

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Abstract

We consider the asymmetric simple exclusion process (ASEP) on the positive integers with an open boundary condition. We show that, when starting devoid of particles and for a certain boundary condition, the height function at the origin fluctuates asymptotically (in large time τ) according to the Tracy–Widom Gaussian orthogonal ensemble distribution on the $\tau^{1/3}$ -scale. This is the first example of Kardar–Parisi–Zhang asymptotics for a half-space system outside the class of free-fermionic/determinantal/Pfaffian models.

Our main tool in this analysis is a new class of probability measures on Young diagrams that we call half-space Macdonald processes, as well as two surprising relations. The first relates a special (Hall–Littlewood) case of these measures to the half-space stochastic six-vertex model (which further limits to the ASEP) using a Yang–Baxter graphical argument. The second relates certain averages under these measures to their half-space (or Pfaffian) Schur process analogues via a refined Littlewood identity.

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1. Introduction

The large-scale statistics of random complex systems are often qualitatively independent from much of the microscopic details of the system, so that the laws of large universality classes can be probed via exactly solvable examples. In particular, particle systems in one spatial dimension—modeling, for instance, phenomena in nonequilibrium transport, traffic jams, and interface growth models—are believed to lie, under mild hypotheses, in the Kardar–Parisi–Zhang (KPZ) universality class (see [28], [37], [38], [54]). The large-scale statistics of such one-dimensional particle systems have been extensively studied in infinite volume. The case of particle systems connected to boundary reservoirs is physically relevant (see [53]) yet less mathematically tractable.

In this article, we study the asymmetric simple exclusion process (ASEP) on the positive integers with an open boundary at the origin in contact with a reservoir of particles kept at a constant density. It is expected (due to known results for the totally asymmetric simple exclusion process [TASEP], a degeneration of the ASEP) that a phase transition happens depending on the local density imposed by the boundary reservoir at the origin between a maximal-current phase and a low-density phase. The critical case happens when the boundary imposes an average particle density equal to $1/2$ at the origin. In this article, we study the statistics of the number of particles in the system when started empty and with boundary conditions tuned to this critical point. We prove that, after a very long time τ , the random variable scales (around its law of large numbers centering) like $\tau^{1/3}$ and converges in this scale weakly to the GOE Tracy–Widom distribution (Theorem A).¹ Further, our results also shed light on the distribution of the solution to the KPZ equation with Neumann boundary condition (Theorem B), which arises as a limit of the height function of the weakly asymmetric half-line ASEP around this critical point (see [31], [46]).

This is the first proof of (KPZ/random-matrix-theoretic) asymptotics in a non-free-fermionic half-space model. Free-fermionic full-space systems have been well studied via robust mathematical approaches, such as the Schur processes (see [44]). These are determinantal systems, meaning that correlation functions are written as determinants of a large matrix. The half-space analogues of such systems are Pfaffian Schur processes (see [6], [23], [47], [49]), whose correlation functions are given via

¹ $G(U/O/S)E$ stands for *Gaussian (unitary/orthogonal/symplectic) ensemble*, and Tracy–Widom distributions were introduced in [55] and [56] as the limiting distributions of the fluctuations of the largest eigenvalues of these ensembles. See Definition 6.1 in the GOE case.

Pfaffians. The full- and half-space TASEP (where jumps only go in one direction) and a small handful of other models fit into the free-fermionic framework.

The ASEP and many other important models do not fit into the free-fermionic framework. In the last decade, starting from the work of Tracy and Widom [58] on the ASEP (on the full line), many KPZ-type limit theorems have been obtained for non-free-fermionic models in a full space. These results have helped refine and expand the notion of KPZ universality. Some attempts have been made to study similar half-space systems, but until now no method has yielded rigorous distributional asymptotics without a Pfaffian structure. Among the existing works on non-free-fermionic half-space systems, O’Connell, Seppäläinen, and Zygouras [43] studied the log-gamma directed polymer in a half-quadrant using properties of the geometric Robinson–Schensted–Knuth algorithm on symmetric matrices, and they conjectured integral formulas, but these are presently not amenable for asymptotic analysis. Using the coordinate Bethe ansatz, Tracy and Widom [59] derived integral formulas for the transition probabilities in the half-line ASEP for certain specific boundary conditions, but these formulas are not amenable to asymptotic analysis either.

Inspired by recent developments relating the integrability of the ASEP on the full line to the stochastic six-vertex model (see [3], [19]–[21]), we study the half-line ASEP as a scaling limit of a stochastic six-vertex model in a half-quadrant with a boundary condition corresponding to off-diagonally symmetric alternating-sign matrices considered in [40]. Our analysis of the half-quadrant stochastic six-vertex model relies upon two surprising relations.

The first relation is between the half-quadrant stochastic six-vertex model and a family of measures on sequences of partitions that we call the *half-space Hall–Littlewood processes*. These measures (see Definition 2.2) are half-space variants of Macdonald processes (introduced in [17] for the full-space case) that generalize Pfaffian Schur processes by replacing Schur functions by Macdonald symmetric functions which rely on two parameters, q , t . Using a graphical interpretation of the Yang–Baxter and reflection equations, we show that the height function in the half-quadrant stochastic six-vertex model has the same law as an observable of the half-space Hall–Littlewood process, that is, the degeneration of the half-space Macdonald process for $q = 0$ (Theorem 4.4).

The second relation is between certain expectations of observables under the half-space Hall–Littlewood and Schur processes. Extracting statistical information from half-space Macdonald processes is, in general, a difficult task (see [8] for an approach using Macdonald operators in the spirit of [17]). In this article, an important technical tool that will considerably simplify our analysis is a refined Littlewood summation identity (Proposition 2.7) for Macdonald symmetric polynomials, which was conjectured in [11] and proved in [48]. This allows us to relate certain observables of Mac-

donald measures for different values of q and consequently connect the Pfaffian Schur process (case $q = t$) and the half-space Hall–Littlewood process (case $q = 0$). The outcome is finally an identity between the t -deformed Laplace transform of the current in the half-line ASEP and a multiplicative functional of a Pfaffian point process with an explicit correlation kernel (Proposition 5.4), which can be analyzed (using Pfaffian point process methods) asymptotically in several interesting limit regimes.

In statistical mechanics, a lot of effort went into obtaining determinant representations for partition functions or correlation functions for the XXZ spin chain/six-vertex model, despite the fact that the model is non-free-fermionic (see, e.g., the review [39]). Our approach achieves this general goal for the half-line ASEP and the half-space stochastic six-vertex model: we uncover a “hidden” fermionic structure and hence compute observables as Pfaffians.

ASEP with an open boundary

The ASEP on a half-line with an open boundary at the origin is an interacting particle system on $\mathbb{Z}_{>0}$ where each site is occupied by at most one particle. Formally, this is a continuous-time Markov process on the state of particle configurations (see Definition 5.1). Each particle jumps by one to the right at rate p and to the left at rate q , with $q < p$, provided that the target site is empty. At the origin, we have a reservoir of particles that injects a particle at site 1 (whenever it is empty) at rate α and removes a particle from site 1 (whenever it is occupied) at rate γ (see Figure 1).

It was proved in [41] that, when

$$\frac{\alpha}{p} + \frac{\gamma}{q} = 1, \tag{1.1}$$

there exist stationary measures for this process. Moreover, by assuming (1.1), there is a phase transition as $\varrho = \frac{\alpha}{p}$ varies. This parameter $\varrho \in (0, 1)$ corresponds to the density of particles that the reservoir imposes at site 1. When $\varrho < 1/2$, the system admits stationary measures that are product measures such that each site is occupied independently with probability ϱ . When $\varrho \geq 1/2$, stationary measures are spatially correlated and more complicated: there is a rarefaction fan with a density of particles ϱ near the origin and density $1/2$ at $+\infty$.

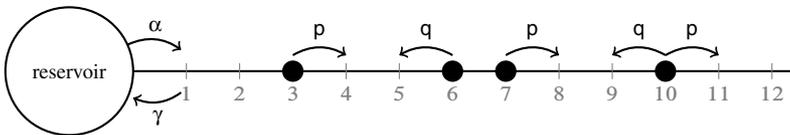


Figure 1. Jump rates in the half-line ASEP. In this article we will study more precisely the case when $p = 1$, $q = t$, $\alpha = 1/2$, $\gamma = t/2$.

When $\mathbf{q} = 0$, the process becomes the TASEP, and this phase transition is much better understood. It was shown in [7] and [5] (see also [49]) that, by starting from an empty configuration, the total number of particles in the system $N(\tau)$ at time τ has Gaussian fluctuations on the scale $\tau^{1/2}$ when $\rho < 1/2$, but has Tracy–Widom GSE fluctuations on the scale $\tau^{1/3}$ when $\rho > 1/2$ and has Tracy–Widom GOE fluctuations on the scale $\tau^{1/3}$ in the critical case $\rho = 1/2$. We refer to [5, Section 6.1] for a heuristic explanation of this phase transition using last passage percolation.

Let us go back to the asymmetric case. We expect that, modulo a rescaling of time by $\mathbf{p} - \mathbf{q}$, the total number of particles in the half-line ASEP undergoes the same phase transition with the same limiting statistics (based on the fact that the full-line TASEP and ASEP asymptotics are identical modulo such time rescaling). Before stating our results, let us mention some of the progress made to uncover the integrability of the model. The stationary distributions in the open ASEP with one or two boundaries can be computed via the matrix product ansatz (see [32], [35]). This realization led to a number of results (generally in the physics literature) such as the derivation of the hydrodynamic limit, understanding of phase diagrams, and large deviation principles. There exists an abundant literature on the subject (see for instance [34] and references therein). An alternative understanding of stationary measures for the half-line ASEP was proposed in [51] using the staircase tableaux (see also [25]–[27], [62]). For the half-line ASEP with a finite constant number of particles (i.e., with closed boundary conditions $\alpha = \gamma = 0$), Tracy and Widom [60] were able to express transition probabilities using the coordinate Bethe ansatz. For the half-line ASEP with general boundary condition, Tracy and Widom [59] derived integral formulas for the transition probabilities, combining the formulas in the closed boundary case with the analysis of the reservoir. These formulas are explicit only when α or γ equal 0 and are presently not amenable for asymptotic analysis in any case.

Let us now state our main result. Without loss of generality,² we can assume that $\mathbf{p} = 1$. The left jump rate \mathbf{q} will be denoted t , as it will coincide in our analysis with the deformation parameter t of Macdonald symmetric functions. Hence, we will denote time rather by the letter τ or T . Our main result is a limit theorem about the fluctuations of the current in the ASEP in the critical case.

THEOREM A (Theorem 6.2)

Under the assumption (1.1) (existence of stationary measures) and for $\rho = 1/2$ (critical density), that is, when the jump rates are given by

$$\mathbf{p} = 1, \quad \mathbf{q} = t, \quad \alpha = 1/2, \quad \gamma = t/2,$$

²Multiplying all jump parameters by a constant corresponds to a division of time by the same multiplicative factor.

we have for any $t \in [0, 1)$ the weak convergence

$$\frac{\frac{T}{4} - N\left(\frac{T}{1-t}\right)}{2^{-4/3}T^{1/3}} \xrightarrow{T \rightarrow +\infty} \mathcal{L}_{\text{GOE}},$$

where $N(\tau)$ denotes the total number of particles in the half-line ASEP at time τ and \mathcal{L}_{GOE} is the Tracy–Widom GOE distribution (see Definition 6.1).

When $t = 0$, the ASEP becomes the TASEP, and the result was proved as Theorem 1.3 in [5] (see also [7], [49] for very similar results in the context of last passage percolation in a half-space with geometric weights). For the half-line TASEP, the current fluctuations are also known for $\varrho \neq 1/2$ (equivalently, $\alpha \neq 1/2$; see [5]). Thus, we expect that current fluctuations of the half-line open ASEP are Tracy–Widom GSE distributed when $\varrho > 1/2$ and Gaussian when $\varrho < 1/2$. Extending Theorem A to other values of ϱ does not seem to be immediately accessible from the techniques developed in this article.

Stochastic six-vertex model in a half-quadrant

We will approach the half-line ASEP through a scaling limit of another integrable model, the stochastic six-vertex model in a half-quadrant, which we believe is also interesting in its own right. On the whole line \mathbb{Z} , this approach was recently used for studying the ASEP in [2], [3], [21], and [29]. Our half-space model is closely related to off-diagonally symmetric alternating-sign matrices, whose weighted enumeration was computed in [40]. Consider the half-quadrant $\{(x, y) \in \mathbb{Z}_{>0}^2 : x \leq y\}$. The stochastic six-vertex model in the half-quadrant is a probability measure on collections of up-right paths (see Figure 2) determined by the Boltzmann weights (see Section 4.1 for a more precise definition)

$$\begin{aligned} \mathbb{P}\left(\begin{array}{c} \vdots \\ \dashrightarrow \\ \vdots \end{array}\right) &= \frac{1 - a_x a_y}{1 - t a_x a_y}, & \mathbb{P}\left(\begin{array}{c} \vdots \\ \dashrightarrow \\ \vdots \end{array}\right) &= \frac{(1-t)a_x a_y}{1 - t a_x a_y}, \\ \mathbb{P}\left(\begin{array}{c} \vdots \\ \vdots \\ \vdots \end{array}\right) &= \frac{t(1 - a_x a_y)}{1 - t a_x a_y}, & \mathbb{P}\left(\begin{array}{c} \vdots \\ \vdots \\ \vdots \end{array}\right) &= \frac{1-t}{1 - t a_x a_y}, \end{aligned}$$

and boundary condition

$$\mathbb{P}\left(\begin{array}{c} \vdots \\ \dashrightarrow \\ \vdots \end{array}\right) = \mathbb{P}\left(\begin{array}{c} \vdots \\ \dashrightarrow \\ \vdots \end{array}\right) = 1, \quad \mathbb{P}\left(\begin{array}{c} \vdots \\ \dashrightarrow \\ \vdots \end{array}\right) = \mathbb{P}\left(\begin{array}{c} \vdots \\ \dashrightarrow \\ \vdots \end{array}\right) = 0. \quad (1.2)$$

Since these weights are stochastic, in the sense that

$$\mathbb{P}\left(\begin{array}{c} \vdots \\ \dashrightarrow \\ \vdots \end{array}\right) + \mathbb{P}\left(\begin{array}{c} \vdots \\ \dashrightarrow \\ \vdots \end{array}\right) = \mathbb{P}\left(\begin{array}{c} \vdots \\ \dashrightarrow \\ \vdots \end{array}\right) + \mathbb{P}\left(\begin{array}{c} \vdots \\ \dashrightarrow \\ \vdots \end{array}\right) = 1,$$

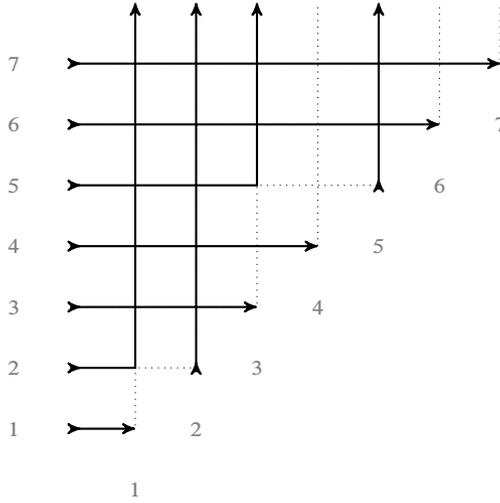


Figure 2. Sample configuration of the stochastic six-vertex model in a half-quadrant.

the measure on paths can be constructed in a Markovian way starting from the left boundary of the quadrant. When the parameters a_x go to 1, the paths will turn at almost every vertex point and, hence, follow a straight staircase path in the $\pi/4$ -direction. If one scales those parameters as $a_x \equiv 1 - \varepsilon$ and rescales time by ε^{-1} , one can interpret the horizontal positions of paths in a finite neighborhood of the diagonal as a particle system that converges to the half-line ASEP as ε goes to 0 (see Proposition 5.2). The assumption that $q = 1/2$ in Theorem A comes from the specific choice of boundary condition (1.2). More general boundary conditions do not seem to be related to the half-space Hall–Littlewood measure considered in this article. It is likely that a result similar to Theorem A can also be proved for the six-vertex model in a half-quadrant by our methods (see Remark 4.15), but we do not pursue that here.

KPZ equation on the positive reals

The KPZ equation on $\mathbb{R}_{\geq 0}$ with Neumann boundary condition is the (ill-posed) stochastic partial differential equation

$$\begin{cases} \partial_\tau \mathcal{H} = \frac{1}{2} \Delta \mathcal{H} + \frac{1}{2} (\partial_x \mathcal{H})^2 + \dot{W}, \\ \partial_x \mathcal{H}(\tau, x)|_{x=0} = A \quad (\forall \tau > 0), \end{cases} \tag{1.3}$$

where W is a space-time white noise. We say that \mathcal{H} solves this equation in the Cole–Hopf sense with narrow-wedge initial condition when $\mathcal{H} = \log \mathcal{L}$ and \mathcal{L} is a mild solution (see Definition 7.1) to the multiplicative stochastic heat equation with Robin boundary condition

$$\begin{cases} \partial_\tau \mathcal{Z} = \frac{1}{2} \Delta \mathcal{Z} + \mathcal{Z} \dot{W}, \\ \partial_x \mathcal{Z}(\tau, x)|_{x=0} = A \mathcal{Z}(\tau, 0) \quad (\forall \tau > 0), \end{cases} \quad (1.4)$$

with delta initial condition. Based on the convergence of the weakly asymmetric ASEP to the KPZ equation from [31], we expect that a certain limit of the half-line ASEP height function (denoted $\mathcal{H}(\tau)$ in the next theorem) weakly converges as t goes to 1 and has the same distribution as $\mathcal{H}(\tau, 0)$, where \mathcal{H} solves (1.3) with $A = -1/2$. We explain, however, in Section 7 that the results of [31] do not directly apply to the boundary and initial conditions that we are considering ([31] assumes $A \geq 0$ and Hölder continuous initial conditions). After the posting of the first version of this article, [46, Theorem 1.2] extended the convergence result from [31] and showed that the random variable denoted $\mathcal{H}(\tau)$ in the next theorem has the same distribution as $\mathcal{H}(\tau, 0)$, where \mathcal{H} solves (1.3) with $A = -1/2$.

THEOREM B (Theorem 7.6 and Corollary 7.7)

Under the scalings

$$t = e^{-\epsilon}, \quad \tau = \frac{\epsilon^{-3} \tilde{\tau}}{1-t} \approx \epsilon^{-4} \tilde{\tau},$$

the random variable

$$\mathcal{U}_\epsilon(\tilde{\tau}) = \frac{t^{(N(\tau) - \epsilon^{-3} \tilde{\tau}/4)}}{1-t^2}$$

weakly converges as $\epsilon \rightarrow 0$ to a positive random variable $\mathcal{U}(\tilde{\tau})$. Furthermore, if

$$\mathcal{H}(\tau) := \log(4 \mathcal{U}(8\tau)) - \frac{\tau}{24},$$

then for any $z > 0$,

$$\mathbb{E} \left[\exp \left(\frac{-z}{4} \exp \left(\frac{\tau}{24} + \mathcal{H}(\tau) \right) \right) \right] = \mathbb{E} \left[\prod_{i=1}^{+\infty} \frac{1}{\sqrt{1 + z \exp((\tau/2)^{1/3} \mathfrak{a}_i)}} \right],$$

where $\{\mathfrak{a}_i\}_{i=1}^{\infty}$ forms the GOE point process (i.e., the sequence of rescaled eigenvalues of a matrix from the GOE; see Definition 6.1).

Remark 1.1

One can deduce immediately from the above theorem that, as τ goes to infinity,

$$\lim_{\tau \rightarrow +\infty} \mathbb{P} \left(\frac{\mathcal{H}(\tau) + \frac{\tau}{24}}{2^{-1/3} \tau^{1/3}} \leq x \right) = \mathbb{P}(\mathfrak{a}_1 \leq x) = F_{\text{GOE}}(x).$$

This is the half-space analogue of [4, Corollary 1.3] (see also [24], [33], [50]), where a similar limit theorem was proved for the solution to the KPZ equation on \mathbb{R} . (The

scaling there is exactly the same, but the limit distribution is the Tracy–Widom GUE distribution instead of the GOE.)

Remark 1.2

The KPZ equation (1.3) with boundary parameter A is considered in the case $A = +\infty$ in the physics paper [36], where large-time Tracy–Widom GSE asymptotics are obtained via a nonrigorous replica method. The paper [15] studies the case $A = 0$, though the results are also based on a nonrigorous replica method and some partially conjectural combinatorial simplifications. It is not yet clear whether the conjectural results from the latter work are compatible with the assumption that the ASEP and TASEP have the same fluctuations. Indeed, if the height function in the ASEP and TASEP satisfies the same limit theorem modulo a rescaling of time by the asymmetry, one would expect that, when the parameters of the ASEP are scaled so as to obtain $A = 0$ in the KPZ equation limit, the large-time fluctuations of the height function would be related to a crossover distribution between Tracy–Widom GOE and GSE distributions as in [5] and [7] (rather than the Tracy–Widom GSE distribution in [15], which would arise only when the density of particles enforced by the boundary near the origin is strictly greater than $1/2$).

Outline of the article

In Section 2, we define half-space Macdonald processes and explain how a refined Littlewood summation identity for Macdonald symmetric polynomials (see Proposition 2.7) allows us to relate certain observables of half-space Macdonald measures for different values of q . More precisely, in Section 3 we express certain observables of the half-space Hall–Littlewood process (case $q = 0$) as Fredholm Pfaffians involving the correlation kernel of the Pfaffian Schur process (case $q = t$). In Section 4, we show that half-space Hall–Littlewood processes are related to the stochastic six-vertex model in a half-quadrant, horizontal sections of the latter being marginals of the former. This type of connection between Macdonald measures and a (higher-spin) vertex model goes back to [14], but our proof is a half-space variant of the corresponding full-space result in [16]. The main ingredients are a t -boson representation of Hall–Littlewood polynomials and also graphical interpretations of the Yang–Baxter and reflection equations from [63]. Under a scaling limit, the height function of the stochastic six-vertex model converges to the ASEP, so that we obtain in Section 5 a Fredholm Pfaffian formula characterizing the distribution of the current in the half-line ASEP. We exploit this formula in two asymptotic regimes. In Section 6, we perform asymptotic analysis on these formulas to prove the convergence of the current in the ASEP to the GOE Tracy–Widom distribution (Theorem 6.2). In Section 7 we discuss the convergence of the ASEP height function to the KPZ equation based

on [31] and prove that the height function of the ASEP at the origin converges in the weak-asymmetry regime to a multiplicative functional of the GOE point process (Theorem 7.6 and Corollary 7.7).

2. Half-space Macdonald processes

A *partition* λ is a nonincreasing sequence of nonnegative integers $\lambda_1 \geq \lambda_2 \geq \dots$ only finitely many of which are nonzero. We will sometimes use the notation $\lambda = 1^{m_1} 2^{m_2} \dots$ for a partition λ , where m_j is the multiplicity of the integer j in the sequence of λ_i 's. We denote by \mathbb{Y} the set of all partitions. The *length* of λ is the number of nonzero elements and is denoted $\ell(\lambda)$. The transpose λ' of a partition is defined by $\lambda'_i = \#\{j : \lambda_j \geq i\}$. In particular, $\lambda'_1 = \ell(\lambda)$. A partition can be identified with a *Young diagram*. For a box \square in a Young diagram, $\ell(\square)$ is equal to the number of boxes in the diagram below it (the leg length) and $a(\square)$ is equal to the number of boxes in the diagram to the right of it (the arm length; see Figure 3). A partition is called *even* if all λ_i 's are even. We write $\mu \subseteq \lambda$ if $\mu_i \leq \lambda_i$ for all i and call λ/μ a *skew Young diagram*. A partition μ *interlaces* with λ (denoted by $\mu < \lambda$) if, for all i , $\lambda_i \geq \mu_i \geq \lambda_{i+1}$. In the language of Young diagrams, this means that λ can be obtained from μ by adding a *horizontal strip* in which at most one box is added per column.

Let \mathbf{Sym} be the ring of symmetric functions in countably many independent variables (see [42, Chapter I, Section 2]). The *skew Macdonald P-functions* (resp., *Q-functions*) $P_{\lambda/\mu}$ (resp., $Q_{\lambda/\mu}$) (introduced in [42, Chapter VI]) are symmetric functions indexed by skew partitions λ/μ and have coefficients that are rational functions of two auxiliary parameters q, t , which we will assume to be in $[0, 1)$ throughout the article. Macdonald symmetric functions become Hall–Littlewood symmetric functions when $q = 0$ and become Schur functions when $q = t$.

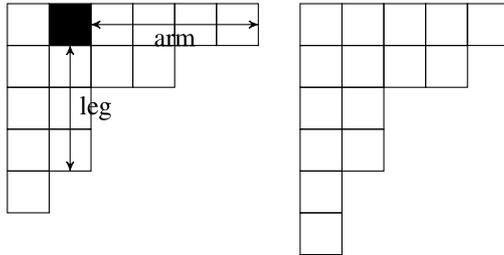


Figure 3. The leftmost diagram corresponds with partition $\lambda = (6, 4, 2, 2, 1)$. The black box has arm length $a(\blacksquare) = 4$ and leg length $\ell(\blacksquare) = 3$. The next diagram is λ 's transpose $\lambda' = (5, 4, 2, 2, 1, 1)$.

For $\lambda \in \mathbb{Y}$, define symmetric functions

$$\mathcal{E}_\lambda = \sum_{\mu' \text{ even}} b_\mu^{\text{el}} Q_{\lambda/\mu}, \quad (2.1)$$

where “el” stands for “even leg” and $b_\mu^{\text{el}} \in \mathbb{Q}[q, t]$ is given by

$$b_\mu^{\text{el}} = \prod_{\substack{\square \in \mu \\ \ell(\square) \text{ even}}} b_\mu(\square), \quad b_\mu(\square) = \begin{cases} \frac{1-q^a t^{\ell+1}}{1-q^{a+1} t^\ell} & \square \in \mu, \\ 1 & \square \notin \mu, \end{cases} \quad (2.2)$$

with $\ell = \ell(\square)$ and $a = a(\square)$ in the definition of $b_\mu(\square)$. All summations over partitions will always be over the set \mathbb{Y} , sometimes subject to specified additional constraints, for example, μ' even.

We now recall or derive certain identities involving Macdonald symmetric functions which will be used in the remainder of the article. The skew Cauchy identity (see [42, Section VI.7]) holds for two sets of formal variables x and y ,

$$\sum_{\kappa} P_{\kappa/v}(x) Q_{\kappa/\lambda}(y) = \Pi(x; y) \sum_{\tau} Q_{v/\tau}(y) P_{\lambda/\tau}(x), \quad (2.3)$$

where $\Pi(x; y)$ is given by (see [42, Chapter VI, (2.5)])

$$\Pi(x; y) := \sum_{\kappa \in \mathbb{Y}} P_\kappa(x) Q_\kappa(y) = \prod_{i,j} \phi(x_i y_j), \quad \text{where } \phi(u) = \frac{(tu; q)_\infty}{(u; q)_\infty} \quad (2.4)$$

and $(a; q)_n = (1-a)(1-qa) \cdots (1-q^{n-1}a)$ is the q -Pochhammer symbol (with infinite product form when $n = \infty$). Macdonald P - and Q -functions satisfy a branching rule whereby (see [42, Section VI.7])

$$\begin{aligned} \sum_{\mu} P_{v/\mu}(x) P_{\mu/\lambda}(y) &= P_{v/\lambda}(x, y) \quad \text{and} \\ \sum_{\mu} Q_{v/\mu}(x) Q_{\mu/\lambda}(y) &= Q_{v/\lambda}(x, y). \end{aligned} \quad (2.5)$$

They also satisfy a Littlewood identity (see [42, Section VI.7, Example 4(i)])

$$\sum_{v' \text{ even}} b_v^{\text{el}} P_v(x) = \prod_{i < j} \phi(x_i x_j) =: \Phi(x). \quad (2.6)$$

Turning to the \mathcal{E}_λ -function, it follows from the definition along with the branching rule (2.5) that

$$\sum_{\mu} Q_{\lambda/\mu}(x) \mathcal{E}_\mu(y) = \mathcal{E}_\lambda(x, y). \quad (2.7)$$

We also have a skew version of the Littlewood identity.

PROPOSITION 2.1

For a set of formal variables x ,

$$\sum_{v' \text{ even}} b_v^{\text{el}} P_{v/\lambda}(x) = \Phi(x) \sum_{\mu' \text{ even}} b_\mu^{\text{el}} Q_{\lambda/\mu}(x) = \Phi(x) \mathcal{E}_\lambda(x). \quad (2.8)$$

Proof

Consider first the one-variable case. $P_{v/\lambda}(x)$ and $Q_{\lambda/\mu}(x)$ are 0 unless $\mu < \lambda < v$. For a given $\lambda \in \mathbb{Y}$, there are unique μ and v such that μ' and v' are even and $\mu < \lambda < v$. It was proved in [42, p. 350, (4)] by an explicit computation that

$$b_v^{\text{el}} P_{v/\lambda}(x) = b_\mu^{\text{el}} Q_{\lambda/\mu}(x).$$

Since in the single-variable case $\Phi(x) = 1$, the identity (2.8) is established in that case. To deduce the result for more general $x = (x_1, \dots, x_n)$, denote $F_\lambda(x) = \sum_{v' \text{ even}} b_v^{\text{el}} P_{v/\lambda}(x)$, so that

$$\begin{aligned} F_\lambda(x_1, \dots, x_n) &= \sum_{v' \text{ even}} b_v^{\text{el}} P_{v/\lambda}(x_1, \dots, x_n) \\ &= \sum_{v' \text{ even}} b_v^{\text{el}} \sum_{\tau} P_{v/\tau}(x_n) P_{\tau/\lambda}(x_1, \dots, x_{n-1}) \\ &= \sum_{\tau} \sum_{\mu' \text{ even}} b_\mu^{\text{el}} Q_{\tau/\mu}(x_n) P_{\tau/\lambda}(x_1, \dots, x_{n-1}) \\ &= \Pi(x_1, \dots, x_{n-1}; x_1) \sum_{\mu' \text{ even}} b_\mu^{\text{el}} \sum_{\tau} Q_{\lambda/\tau}(x_n) P_{\mu/\tau}(x_1, \dots, x_{n-1}) \\ &= \Pi(x_1, \dots, x_{n-1}; x_1) \sum_{\tau} Q_{\lambda/\tau}(x_n) F_\tau(x_1, \dots, x_{n-1}), \end{aligned}$$

where in the second equality we have used the branching rule (2.5), in the third equality we have used the single-variable case established above, and in the fourth equality we have used the skew Cauchy identity (2.3). We may now iterate the relation

$$F_\lambda(x_1, \dots, x_n) = \Pi(x_1, \dots, x_{n-1}; x_1) \sum_{\tau} Q_{\lambda/\tau}(x_n) F_\tau(x_1, \dots, x_{n-1}),$$

use the branching rule (2.5), and deduce that

$$F_\lambda(x) = \prod_{i < j} \phi(x_i x_j) \sum_{\mu' \text{ even}} b_\mu^{\text{el}} Q_{\lambda/\mu}(x),$$

as desired. □

Now combining the skew Cauchy and skew Littlewood identities, we obtain

$$\sum_{\mu} \mathcal{E}_{\mu}(x) P_{\mu/\lambda}(y) = \Pi(x, y) \Phi(y) \mathcal{E}_{\lambda}(x, y). \quad (2.9)$$

In particular,

$$\sum_{\mu} \mathcal{E}_{\mu}(x) P_{\mu}(y) = \Pi(x, y) \Phi(y). \quad (2.10)$$

A *specialization* ρ of \mathbf{Sym} is an algebra homomorphism of $\mathbf{Sym} \rightarrow \mathbb{C}$. For instance, evaluating symmetric functions at a fixed finite set of variables defines such a homomorphism (see [17, Section 2.2.1] for a more detailed discussion). We denote the application of ρ to $f \in \mathbf{Sym}$ as $f(\rho)$, thus extending the usual notation for the evaluation at a set of variables. The *trivial* specialization $\rho = \emptyset$ takes the value 1 for the constant function $1 \in \mathbf{Sym}$ and 0 for all homogeneous functions $f \in \mathbf{Sym}$ of higher degree. The *union* of two specializations ρ_1, ρ_2 is defined via the relation

$$p_k(\rho_1, \rho_2) = p_k(\rho_1) + p_k(\rho_2)$$

and extended to all of \mathbf{Sym} by linearity. Here $p_k(x) = x_1^k + x_2^k + \dots$ are the Newton power sum symmetric functions. Notationally, we may write the union of ρ_1, ρ_2 by putting a comma between them. We say a specialization ρ is *Macdonald nonnegative* if, for every skew diagram λ/μ , $P_{\lambda/\mu}(\rho) \geq 0$.

Definition 2.2

For two Macdonald-nonnegative specializations ρ^+, ρ^- , we define the *half-space Macdonald measure* as a probability measure \mathbb{P}^{MM} on Young diagrams $\lambda \in \mathbb{Y}$ such that

$$\mathbb{P}^{\text{MM}}(\lambda) = \frac{1}{\Pi(\rho^+; \rho^-) \Phi(\rho^+)} P_{\lambda}(\rho^+) \mathcal{E}_{\lambda}(\rho^-).$$

We will denote by \mathbb{E}^{PMM} the corresponding expectation. This is a well-defined probability measure thanks to (2.10), provided that the series converges. In this article, we deal with Macdonald-nonnegative specializations corresponding to evaluating functions into finitely many symmetric variables in $(0, 1)$, so that the sums (2.3) and (2.8) always converge.

Definition 2.3

For Macdonald-nonnegative specializations $\rho_1^+, \dots, \rho_n^+$ and $\rho_1^-, \dots, \rho_n^-$, we define the *half-space Macdonald process* as a probability measure \mathbb{P}^{MP} on sequences of Young diagrams

$$\emptyset \subset \lambda^{(1)} \supset \mu^{(1)} \subset \lambda^{(2)} \supset \mu^{(2)} \subset \dots \supset \mu^{(n-1)} \subset \lambda^{(n)} \supset \emptyset$$

such that

$$\begin{aligned} & \text{PMM}(\lambda, \mu) \\ &= \frac{1}{Z(\rho)} P_{\lambda^{(1)}}(\rho_1^+) Q_{\lambda^{(1)}/\mu^{(1)}}(\rho_1^-) \\ & \quad \times P_{\lambda^{(2)}/\mu^{(1)}}(\rho_2^+) Q_{\lambda^{(2)}/\mu^{(2)}}(\rho_2^-) \cdots P_{\lambda^{(n)}/\mu^{(n-1)}}(\rho_n^+) \mathcal{E}_{\lambda^{(n)}}(\rho_n^-), \end{aligned} \quad (2.11)$$

where

$$Z(\rho) = \Phi(\rho^+) \prod_{i \leq j} \Pi(\rho_i^+; \rho_j^-)$$

and $\rho^+ = (\rho_1^+, \dots, \rho_N^+)$. We will denote by \mathbb{E}^{PMP} the corresponding expectation. One may check that this indeed defines a probability measure by repeated application of the skew Cauchy and Littlewood identities (2.3) and (2.8), provided that all series converge.

PROPOSITION 2.4

Under the notations of Definition 2.2 and 2.3, the marginal distribution of $\lambda^{(k)}$ under the half-space Macdonald process is the half-space Macdonald measure with specializations

$$\rho^+ = (\rho_1^+, \dots, \rho_k^+), \quad \rho^- = (\rho_k^-, \dots, \rho_n^-).$$

Proof

The result follows from summing the probabilities (2.11) over all $\mu^{(i)}$ and $\lambda^{(j)}$ for $j \neq k$, using the branching rule (2.5), the skew Cauchy identity (2.3), and the summation formula (2.9). \square

Remark 2.5

Half-space Macdonald measures/processes naturally generalize Pfaffian Schur measures/processes introduced in [23] and [6], the latter corresponding to the degeneration when $q = t$. The term *Pfaffian* comes from the fact that the Pfaffian Schur process defines a Pfaffian point process (see [23, Theorem 3.3]). However, half-space Macdonald processes do not correspond to any Pfaffian point process.

Definition 2.6

We define the *ascending half-space Macdonald process* as the half-space Macdonald process defined by specializations³ $\rho_1^- = \dots = \rho_n^- = \emptyset$ and $\rho_i^+ = (a_i)$ (specialization

into a single variable a_i), where the variables (a_1, \dots, a_n) are in $(0, 1)$. This process is supported on sequences of interlaced partitions

$$\lambda^{(1)} \prec \lambda^{(2)} \prec \dots \prec \lambda^{(n)}.$$

In the rest of this section, we focus on the marginal of $\lambda^{(n)}$ for the ascending half-space Macdonald process, that is, the half-space Macdonald measure with specializations

$$\rho^+ = (a_1, \dots, a_n), \quad \rho^- = \emptyset.$$

We will need one more identity, first conjectured in [11] and proved in [48].

PROPOSITION 2.7 ([48, Proposition 6.26])

For an even integer n and any $u \in \mathbb{C}$, Macdonald symmetric polynomials satisfy

$$\begin{aligned} \frac{1}{\Phi(x)} \sum_{\lambda' \text{ even}} \prod_{i \text{ even}} (1 - uq^{\lambda_i} t^{n-i}) b_\lambda^{\text{el}} P_\lambda(x_1, \dots, x_n) \\ = \frac{\text{Pf}\left[\frac{x_i - x_j}{1 - x_i x_j} - u \frac{x_i - x_j}{1 - t x_i x_j}\right]}{\text{Pf}\left[\frac{x_i - x_j}{1 - x_i x_j}\right]}. \end{aligned} \quad (2.12)$$

The most striking fact in the above identity (2.12) is that the right-hand side does not depend on q . This yields identities that relate half-space Macdonald measures for different values of q . Consider a set of parameters $a_1, \dots, a_n \in (0, 1)$ for n even, and let us denote by \mathbb{P}^{SM} and \mathbb{E}^{PSM} the half-space Macdonald measure and expectation associated with parameters a_i in the Schur case $q = t$. It is important to note that \mathbb{P}^{SM} and \mathbb{E}^{PSM} do not depend on the parameter $q = t$. In this context, (2.12) implies that, for any $q, t \in [0, 1)$ and $u \in \mathbb{C}$,

$$\mathbb{E}^{\text{PMM}}\left[\prod_{i \text{ even}} (1 - uq^{\lambda_i} t^{n-i})\right] = \mathbb{E}^{\text{PSM}}\left[\prod_{i \text{ even}} (1 - ut^{\lambda_i + n-i})\right]. \quad (2.13)$$

Letting $u = -t^x$ and dividing both sides by $(-t^x; t^2)_\infty$, we rewrite this identity as

$$\mathbb{E}^{\text{PMM}}\left[\prod_{i \in 2\mathbb{Z}_{\geq 0}} \left(\frac{1 + q^{\lambda_{n-i}} t^{x+i}}{1 + t^{x+i}}\right)\right] = \mathbb{E}^{\text{PSM}}\left[\prod_{i \in 2\mathbb{Z}_{\geq 0}} \left(\frac{1 + t^{x+\lambda_{n-i}+i}}{1 + t^{x+i}}\right)\right], \quad (2.14)$$

for any $x \in \mathbb{R}$, with the convention that $\lambda_{-m} = +\infty$ for $m > 0$.

³We could allow ρ_n^- to be arbitrary to study more general boundary conditions.

Remark 2.8

One could prove using (2.14) a half-space analogue of [14, Corollary 5.9]: the distribution of the length of a half-space Macdonald random partition is asymptotically the same whatever the values of q and t , in the sense of asymptotic equivalence as in [14, Definition 5.2]. We do not need this result for the present article.

3. Fredholm Pfaffian formulas*3.1. Notation*

The Pfaffian of a skew-symmetric $2k \times 2k$ matrix A is defined by

$$\text{Pf}(A) = \frac{1}{2^k k!} \sum_{\sigma \in \mathfrak{S}_{2k}} \text{sgn}(\sigma) a_{\sigma(1)\sigma(2)} a_{\sigma(3)\sigma(4)} \cdots a_{\sigma(2k-1)\sigma(2k)}, \quad (3.1)$$

where $\text{sgn}(\sigma)$ is the signature of the permutation σ . Let (\mathbb{X}, μ) be a measure space. For a 2×2 matrix-valued skew-symmetric kernel

$$\mathbf{K}(x, y) = \begin{pmatrix} \mathbf{K}_{11}(x, y) & \mathbf{K}_{12}(x, y) \\ \mathbf{K}_{21}(x, y) & \mathbf{K}_{22}(x, y) \end{pmatrix}, \quad x, y \in \mathbb{X},$$

we define the Fredholm Pfaffian (introduced in [47, Section 8]) by

$$\begin{aligned} & \text{Pf}[\mathbf{J} + \mathbf{K}]_{L^2(\mathbb{X}, \mu)} \\ &= 1 + \sum_{k=1}^{\infty} \frac{1}{k!} \int_{\mathbb{X}} \cdots \int_{\mathbb{X}} \text{Pf}(\mathbf{K}(x_i, x_j))_{i,j=1}^k d\mu^{\otimes k}(x_1, \dots, x_k), \end{aligned} \quad (3.2)$$

provided that the series converges, where $\mu^{\otimes k}$ is the product measure. The kernel \mathbf{J} is defined by

$$\mathbf{J}(x, y) = \delta_{x=y} \begin{pmatrix} 0 & 1 \\ -1 & 0 \end{pmatrix}.$$

For a function $f: \mathbb{X} \rightarrow \mathbb{R}$, we define

$$\begin{aligned} \text{Pf}[\mathbf{J} + f \cdot \mathbf{K}]_{L^2(\mathbb{X}, \mu)} &:= \text{Pf}[\mathbf{J} + \mathbf{K}]_{L^2(\mathbb{X}, f\mu)} \\ &= 1 + \sum_{k=1}^{\infty} \frac{1}{k!} \int_{\mathbb{X}} \cdots \int_{\mathbb{X}} \left(\prod_{i=1}^k f(x_i) \right) \\ &\quad \times \text{Pf}(\mathbf{K}(x_i, x_j))_{i,j=1}^k d\mu^{\otimes k}(x_1, \dots, x_k), \end{aligned}$$

provided that the series converge. We will use the notation $\text{Pf}[\mathbf{J} + \mathbf{K}]_{L^2(\mathbb{X})}$ when \mathbb{X} is a subset of \mathbb{R}^n equipped with Lebesgue measure, and we will use the notation

$\text{Pf}[\mathbf{J} + \mathbf{K}]_{\ell^2(\mathbb{X})}$ when \mathbb{X} is a discrete set equipped with the counting measure. In order to study limits of Fredholm Pfaffians, we will need that the expansion in (3.2) is absolutely convergent, and for that we will use Hadamard's bound in the form of the next lemma.

LEMMA 3.1 ([5, Lemma 2.5])

Let $\mathbf{K}(x, y)$ be a 2×2 matrix-valued skew-symmetric kernel. Assume that there exist constant $C > 0$ and constants $0 \leq a < b$ such that, for $x, y \in \mathbb{X} \subset \mathbb{R}$,

$$\begin{aligned} |\mathbf{K}_{11}(x, y)| &< C e^{ax+ay}, \\ |\mathbf{K}_{12}(x, y)| &= |\mathbf{K}_{21}(y, x)| < C e^{ax-by}, \\ |\mathbf{K}_{22}(x, y)| &< C e^{-bx-by}. \end{aligned}$$

Then, for all $k \in \mathbb{Z}_{>0}$, $x_1, \dots, x_k \in \mathbb{X}$,

$$|\text{Pf}(\mathbf{K}(x_i, x_j))_{i,j=1}^k| < (2k)^{k/2} C^k \prod_{i=1}^k e^{-(b-a)x_i}.$$

3.2. Pfaffian Schur measure correlation kernel

Consider a Pfaffian Schur measure with parameters $(a_1, \dots, a_n) \in (0, 1)^n$, that is, a measure of the form

$$\text{PSM}(\lambda) = \prod_{i < j} (1 - a_i a_j) s_\lambda(a_1, \dots, a_n) \mathbb{1}_{\lambda' \text{ even}}, \quad \lambda \in \mathbb{Y}.$$

We know from [23, Theorem 3.3] that the random point configuration $\Lambda := \{\lambda_i - i\}_{i \in \mathbb{Z}_{>0}}$ generates a Pfaffian point process on \mathbb{Z} . This means that, for $y_1, \dots, y_k \in \mathbb{Z}$,

$$\mathbb{P}(\{y_1, \dots, y_k\} \subset \Lambda) = \text{Pf}[\mathbf{K}(y_i, y_j)]_{i,j=1}^k,$$

where $\mathbf{K}(u, v)$ is a 2×2 matrix-valued skew-symmetric kernel

$$\mathbf{K}(u, v) = \begin{pmatrix} \mathbf{K}_{11}(u, v) & \mathbf{K}_{12}(u, v) \\ \mathbf{K}_{21}(u, v) & \mathbf{K}_{22}(u, v) \end{pmatrix}, \quad u, v \in \mathbb{Z},$$

with $\mathbf{K}_{12} = -(\mathbf{K}_{21})^T$. We refer to [5, Section 4.1] for general background on Pfaffian point processes. For the Pfaffian Schur process, the kernel is (see [23, Theorem 3.3])

$$\begin{aligned} \mathbf{K}_{11}^{\text{Schur}}(u, v) &= \frac{1}{(2i\pi)^2} \iint \frac{z-w}{(z^2-1)(w^2-1)(zw-1)} \\ &\quad \times \prod_{j=1}^n \left(\frac{(z-a_j)(w-a_j)}{zw(1-a_jz)(1-a_jw)} \right) \frac{dz}{z^u} \frac{dw}{w^v}, \end{aligned}$$

$$\begin{aligned}
\mathbf{K}_{12}^{\text{Schur}}(u, v) &= \frac{1}{(2i\pi)^2} \iint \frac{z-w}{(z^2-1)w(zw-1)} \\
&\quad \times \prod_{j=1}^n \left(\frac{(z-a_j)(w-a_j)}{zw(1-a_jz)(1-a_jw)} \right) \frac{dz}{z^u} \frac{dw}{w^v}, \\
\mathbf{K}_{22}^{\text{Schur}}(u, v) &= \frac{1}{(2i\pi)^2} \iint \frac{z-w}{zw(zw-1)} \\
&\quad \times \prod_{j=1}^n \left(\frac{(z-a_j)(w-a_j)}{zw(1-a_jz)(1-a_jw)} \right) \frac{dz}{z^u} \frac{dw}{w^v},
\end{aligned}$$

where for $\mathbf{K}_{11}^{\text{Schur}}$ and $\mathbf{K}_{12}^{\text{Schur}}$, the contours for z and w are positively oriented circles around 0 with radius between 1 and $\min\{a_j^{-1}\}$ (so that, in particular, $|zw| > 1$ along these contours), whereas for $\mathbf{K}_{22}^{\text{Schur}}$, the contours for z and w are positively oriented circles around 0 with radius smaller than 1 (so that $|zw| < 1$ along these contours).

Note that $-\ell(\lambda)$ is the leftmost hole in the point process Λ (or leftmost point in $\Lambda^{\mathbb{G}} := \mathbb{Z} \setminus \Lambda$). The correlation kernel of $\Lambda^{\mathbb{G}}$ is $\mathbf{K}' = \mathbf{J} - \mathbf{K}^{\text{Schur}}$. Indeed, the probability that a set $Y = \{y_1, \dots, y_k\}$ is included in $\Lambda^{\mathbb{G}}$ is also the probability that there are no points of Λ in Y , that is, the gap probability, which is given by $\text{Pf}[\mathbf{J} - \mathbf{K}^{\text{Schur}}]_{\ell^2(Y)}$. For a finite set Y , the Fredholm Pfaffian on $\ell^2(Y)$ is simply the Pfaffian of the matrix indexed by elements of Y , that is, $\text{Pf}[\mathbf{K}'(y_i, y_j)]_{i,j=1}^k$, such that the correlation kernel of the point process $\Lambda^{\mathbb{G}}$ is \mathbf{K}' as claimed. In particular,

$$\mathbb{P}(-\ell(\lambda) > x) = \text{Pf}[\mathbf{K}^{\text{Schur}}]_{\ell^2(-\infty, x]} = \text{Pf}[\mathbf{J} - \mathbf{K}']_{\ell^2(-\infty, x]}.$$

Shifting the point process by n , we obtain that the correlation kernel of $n + \Lambda^{\mathbb{G}}$ is $\mathbf{K}^{\mathbb{G}}$, where

$$\begin{aligned}
\mathbf{K}_{11}^{\mathbb{G}}(u, v) &= \frac{1}{(2i\pi)^2} \iint \frac{w-z}{(z^2-1)(w^2-1)(zw-1)} \\
&\quad \times \prod_{j=1}^n \left(\frac{(z-a_j)(w-a_j)}{(1-a_jz)(1-a_jw)} \right) \frac{dz}{z^u} \frac{dw}{w^v}, \tag{3.3a}
\end{aligned}$$

$$\begin{aligned}
\mathbf{K}_{12}^{\mathbb{G}}(u, v) &= \mathbb{1}_{u=v} + \frac{1}{(2i\pi)^2} \iint \frac{w-z}{(z^2-1)w(zw-1)} \\
&\quad \times \prod_{j=1}^n \left(\frac{(z-a_j)(w-a_j)}{(1-a_jz)(1-a_jw)} \right) \frac{dz}{z^u} \frac{dw}{w^v}, \tag{3.3b}
\end{aligned}$$

$$\begin{aligned} \mathcal{K}_{22}^{\mathbb{G}}(u, v) &= \frac{1}{(2i\pi)^2} \iint \frac{w-z}{zw(zw-1)} \\ &\quad \times \prod_{j=1}^n \left(\frac{(z-a_j)(w-a_j)}{(1-a_jz)(1-a_jw)} \right) \frac{dz}{z^u} \frac{dw}{w^v}, \end{aligned} \quad (3.3c)$$

where the contours are as before. The point process $n + \Lambda^{\mathbb{G}}$ is almost surely supported on the nonnegative integers, so that one can compute the Fredholm Pfaffian on $\ell^2[0, x]$ instead of $\ell^2(-\infty, x]$. Thus, by [47, Theorem 8.2], one may write that, for any function $f: \mathbb{Z}_{\geq 0} \rightarrow \mathbb{R}$,

$$\mathbb{E}^{\text{PSM}} \left[\prod_{\lambda \in n + \Lambda^{\mathbb{G}}} (1 + f(\lambda)) \right] = \text{Pf}[\mathbf{J} + \mathbf{f} \cdot \mathcal{K}^{\mathbb{G}}]_{\ell^2(\mathbb{Z}_{\geq 0})}, \quad (3.4)$$

whenever both sides admit absolutely convergent expansions. In particular,

$$\text{PSM}(n - \ell(\lambda) > x) = \text{Pf}[\mathbf{J} - \mathcal{K}^{\mathbb{G}}]_{\ell^2[0, x]}.$$

It will be more convenient to work with integral formulas where the contours are all circles with radius less than 1 (because we will later let $a_j \equiv a$ go to 1 and there is a pole at $1/a$). When deforming the contours inside the unit circle, we pick some residues which yield the following formulas.

LEMMA 3.2

Denoting $f(z) = \prod_{j=1}^n \frac{z-a_j}{1-a_jz}$, for n even and $u, v \in \mathbb{Z}_{\geq 0}$, we have that

$$\begin{aligned} \mathcal{K}_{11}^{\mathbb{G}}(u, v) &= \frac{1}{(2i\pi)^2} \iint \frac{(w-z)f(z)f(w)}{(z^2-1)(w^2-1)(zw-1)} \frac{dz}{z^u} \frac{dw}{w^v} \\ &\quad + \frac{1}{2i\pi} \int \frac{f(w)}{w^2-1} \frac{dw}{w^v} \mathbb{1}_{u \in 2\mathbb{Z}} \\ &\quad - \frac{1}{2i\pi} \int \frac{f(w)}{w^2-1} \frac{dw}{w^u} \mathbb{1}_{v \in 2\mathbb{Z}} + r(u, v), \\ \mathcal{K}_{12}^{\mathbb{G}}(u, v) &= \frac{1}{(2i\pi)^2} \iint \frac{(w-z)f(z)f(w)}{(z^2-1)w(zw-1)} \frac{dz}{z^u} \frac{dw}{w^v} \\ &\quad + \frac{1}{2i\pi} \int \frac{f(z)}{z^{v+1}} dz \mathbb{1}_{u \in 2\mathbb{Z}}, \\ \mathcal{K}_{22}^{\mathbb{G}}(u, v) &= \frac{1}{(2i\pi)^2} \iint \frac{(w-z)f(z)f(w)}{zw(zw-1)} \frac{dz}{z^u} \frac{dw}{w^v}, \end{aligned}$$

where the contours are all positively oriented circles around 0 with radius smaller than 1, and

$$r(u, v) = \frac{1}{4}((-1)^u - (-1)^v) + \frac{1}{2} \operatorname{sgn}(v - u) \mathbb{1}_{v-u \in 2\mathbb{Z}+1}, \quad (3.5)$$

with the convention that

$$\operatorname{sgn}(x) = \begin{cases} 1 & \text{if } x > 0, \\ -1 & \text{if } x < 0, \\ 0 & \text{if } x = 0. \end{cases}$$

Proof

Let $\mathcal{C}_{>1}$ be a contour defined by a (positively oriented) circle of radius larger than 1 but arbitrarily close to 1, let $\mathcal{C}_{<1}$ be a contour with radius smaller than 1 but arbitrarily close to 1, and let $\mathcal{C}_{\ll 1}$ be a contour with radius arbitrarily close to 0. In the formula for $\mathbb{K}_{11}^{\mathbb{G}}(u, v)$ in (3.3a), the integration variables z and w are such that $|z|, |w|, |zw| > 1$. We may first deform the z -contour from $\mathcal{C}_{>1}$ to $\mathcal{C}_{\ll 1}$, thus picking residues at $z = \pm 1$ and $z = 1/w$. This yields

$$\begin{aligned} \mathbb{K}_{11}^{\mathbb{G}}(u, v) &= \frac{1}{(2i\pi)^2} \int_{\mathcal{C}_{>1}} dw \int_{\mathcal{C}_{\ll 1}} dz \frac{(w-z)f(z)f(w)}{(z^2-1)(w^2-1)(zw-1)} \frac{1}{z^u} \frac{1}{w^v} \\ &\quad + \frac{1}{2i\pi} \int_{\mathcal{C}_{>1}} dw \frac{1}{2(w^2-1)} \frac{f(w)}{w^v} (f(1) + (-1)^u f(-1)) \\ &\quad + \int_{\mathcal{C}_{>1}} dw \frac{1}{1-w^2} \frac{1}{w^{v-u}}. \end{aligned}$$

Evaluating the residues at ± 1 in the second integral and taking into account that n is even so that $f(1) = f(-1) = 1$, we find

$$\begin{aligned} \mathbb{K}_{11}^{\mathbb{G}}(u, v) &= \frac{1}{(2i\pi)^2} \int_{\mathcal{C}_{>1}} dw \int_{\mathcal{C}_{\ll 1}} dz \frac{(w-z)f(z)f(w)}{(z^2-1)(w^2-1)(zw-1)} \frac{1}{z^u} \frac{1}{w^v} \\ &\quad + \frac{1}{2i\pi} \int_{\mathcal{C}_{\ll 1}} dw \frac{1}{w^2-1} \frac{f(w)}{w^v} \mathbb{1}_{u \in 2\mathbb{Z}} + r(u, v), \end{aligned}$$

where $r(u, v)$ is defined in (3.5). This term $r(u, v)$ corresponds to taking residues in the variable z and then in the variables w in (3.3a) for values of (z, w) equal to

$$\begin{aligned} &(1/w, 1), \quad (1/w, -1), \quad (1/w, 0), \quad (1, 1), \\ &(1, -1), \quad (-1, 1), \quad \text{and} \quad (-1, -1). \end{aligned}$$

One readily checks that the sum of all these residues equals $r(u, v)$.

Deforming the contour for the variable w in the first integral from $\mathcal{C}_{>1}$ to $\mathcal{C}_{\ll 1}$ we pick residues at ± 1 , which yield

$$\begin{aligned}
\mathsf{K}_{11}^{\mathbb{C}}(u, v) &= \frac{1}{(2i\pi)^2} \int_{\mathcal{C}_{\ll 1}} dw \int_{\mathcal{C}_{\ll 1}} dz \frac{(w-z)f(z)f(w)}{(z^2-1)(w^2-1)(zw-1)} \frac{1}{z^u} \frac{1}{w^v} \\
&\quad - \frac{1}{2i\pi} \int_{\mathcal{C}_{\ll 1}} dw \frac{1}{w^2-1} \frac{f(w)}{w^u} \mathbb{1}_{v \in 2\mathbb{Z}} \\
&\quad + \frac{1}{2i\pi} \int_{\mathcal{C}_{\ll 1}} dw \frac{1}{w^2-1} \frac{f(w)}{w^v} \mathbb{1}_{u \in 2\mathbb{Z}} + r(u, v).
\end{aligned}$$

In the formula for $\mathsf{K}_{12}^{\mathbb{C}}(u, v)$ in (3.3b), we may deform the contour for w to $\mathcal{C}_{<1}$ without picking any residue. Then, deforming the contour for z to $\mathcal{C}_{<1}$ we pick residues at $z = 1/w$ and $z = \pm 1$, which yield

$$\begin{aligned}
\mathsf{K}_{12}^{\mathbb{C}}(u, v) &= \mathbb{1}_{u=v} + \frac{1}{(2i\pi)^2} \int_{\mathcal{C}_{<1}} dw \int_{\mathcal{C}_{<1}} dz \frac{(w-z)f(z)f(w)}{(z^2-1)w(zw-1)} \frac{dz}{z^u} \frac{dw}{w^v} \\
&\quad + \frac{1}{2i\pi} \int_{\mathcal{C}_{<1}} dw \frac{-1}{w^{v-u+1}} + \frac{1}{2i\pi} \int_{\mathcal{C}_{<1}} \frac{f(w)}{w^{v+1}} \mathbb{1}_{u \in 2\mathbb{Z}},
\end{aligned}$$

which simplifies to

$$\begin{aligned}
\mathsf{K}_{12}^{\mathbb{C}}(u, v) &= \frac{1}{(2i\pi)^2} \int_{\mathcal{C}_{<1}} dw \int_{\mathcal{C}_{<1}} dz \frac{(w-z)f(z)f(w)}{(z^2-1)w(zw-1)} \frac{dz}{z^u} \frac{dw}{w^v} \\
&\quad + \frac{1}{2i\pi} \int_{\mathcal{C}_{<1}} \frac{f(w)}{w^{v+1}} \mathbb{1}_{u \in 2\mathbb{Z}}. \quad \square
\end{aligned}$$

3.3. Hall–Littlewood observables

We can use the relation between Pfaffian Schur and Hall–Littlewood measure (2.13) to express certain observables of the latter using the correlation kernel of the former.

PROPOSITION 3.3

For any $x \in \mathbb{R}$, $n \in 2\mathbb{Z}_{>0}$, and any parameters $(a_1, \dots, a_n) \in (0, 1)^n$,

$$\mathbb{E}^{\text{PHL}} \left[\frac{1}{(-t^{x+n-\ell(\lambda)}, t^2)_{\infty}} \right] = \text{Pf}[\mathsf{J} + \mathsf{f}_x \cdot \mathsf{K}^{\mathbb{C}}]_{\ell^2(\mathbb{Z}_{\geq 0})}, \quad (3.6)$$

where

$$\mathsf{f}_x(j) = \frac{(-t^{x+j+1}; t^2)_{\infty}}{(-t^{x+j}; t^2)_{\infty}} - 1. \quad (3.7)$$

Before proving this proposition, it is worth noting that $\mathsf{f}_x(j) \in (-1, 0)$ and $\mathsf{f}_x(j)$ is increasing in x and j . This will be of importance later when we consider scaling limits.

Proof

Letting $u = -t^x$ and $q = 0$ in identity (2.13) implies that

$$\mathbb{E}^{\text{PHL}}\left[\prod_{i \in 2\mathbb{Z} \cap [0, n - \ell(\lambda) - 2]} (1 + t^{x+i})\right] = \mathbb{E}^{\text{PSM}}\left[\prod_{i \in 2\mathbb{Z} \cap [0, n]} (1 + t^{x+\lambda_i - i + n})\right]. \quad (3.8)$$

The left-hand side of (3.8) can be rewritten as

$$(-t^x; t^2)_\infty \mathbb{E}^{\text{PHL}}\left[\frac{1}{(-t^{x+n-\ell(\lambda)}, t^2)_\infty}\right],$$

and the right-hand side can be rewritten as

$$\mathbb{E}^{\text{PSM}}\left[\prod_{j \in J} (1 + t^{x+j})\right],$$

where $J = \{\lambda_i - i + n\}_{i \in 2\mathbb{Z} \cup [0, n]}$. Since $\lambda_1 = \lambda_2, \lambda_3 = \lambda_4, \dots$ under the Pfaffian Schur measure that we consider, the set J is characterized by $J \sqcup (J + 1) = (n + \Lambda) \cap \mathbb{Z}_{\geq 0}$, where \sqcup denotes the union of disjoint sets and $\Lambda = \{\lambda_i - i\}_{i \in \mathbb{Z}_{>0}}$ as before. Let us call $P(x)$ the product inside the last expectation. We have

$$P(x)P(x+1) = \prod_{j \in (n+\Lambda) \cap \mathbb{Z}_{\geq 0}} (1 + t^{x+j}) =: N(x).$$

This implies that

$$P(x) = \frac{N(x)N(x+2)N(x+4)\cdots}{N(x+1)N(x+3)N(x+5)\cdots} = \prod_{j \in (n+\Lambda) \cap \mathbb{Z}_{\geq 0}} \frac{(-t^{x+j}; t^2)_\infty}{(-t^{x+j+1}; t^2)_\infty}.$$

Thus, we have shown that

$$\begin{aligned} & \mathbb{E}^{\text{PSM}}\left[\prod_{i \in 2\mathbb{Z} \cap [0, n]} (1 + t^{x+\lambda_i - i + n})\right] \\ &= \mathbb{E}^{\text{PSM}}\left[\prod_{j \in (n+\Lambda) \cap \mathbb{Z}_{\geq 0}} \frac{(-t^{x+j}; t^2)_\infty}{(-t^{x+j+1}; t^2)_\infty}\right]. \end{aligned} \quad (3.9)$$

Note the simplification

$$\prod_{j \in \mathbb{Z}_{\geq 0}} \frac{(-t^{x+j}; t^2)_\infty}{(-t^{x+j+1}; t^2)_\infty} = (-t^x; t^2)_\infty.$$

so that dividing both sides of (3.9) by that quantity gives

$$\begin{aligned} & \frac{1}{(-t^x; t^2)_\infty} \mathbb{E}^{\text{PSM}} \left[\prod_{i \in 2\mathbb{Z} \cap [0, n]} (1 + t^{x+\lambda_i - i + n}) \right] \\ &= \mathbb{E}^{\text{PSM}} \left[\prod_{j \in n + \Lambda^{\mathbb{G}}} \frac{(-t^{x+j+1}; t^2)_\infty}{(-t^{x+j}; t^2)_\infty} \right]. \end{aligned}$$

At this point we have shown the following relation between observables of the half-space Schur and Hall–Littlewood processes:

$$\mathbb{E}^{\text{PHL}} \left[\frac{1}{(-t^{x+n-\ell(\lambda)}, t^2)_\infty} \right] = \mathbb{E}^{\text{PSM}} \left[\prod_{j \in n + \Lambda^{\mathbb{G}}} \frac{(-t^{x+j+1}; t^2)_\infty}{(-t^{x+j}; t^2)_\infty} \right].$$

To complete the proof, we note that the multiplicative functional of the Pfaffian Schur measure on the right-hand side can be computed using (3.4) as in the statement of the proposition. \square

4. Half-space Hall–Littlewood measures and stochastic six-vertex model in a half-quadrant

We will define a measure on lattice paths as in Figure 2. There are two types of vertices: bulk and corner vertices.

4.1. Definition of the model

Consider the square lattice. A *bulk vertex* is the crossing of a horizontal line and a vertical line, as well as the four edges which surround the point of intersection. We refer to the four edges which constitute a vertex as its north, east, south, and west edges, following standard compass orientation. Bulk vertices have the following generic form:

$$\begin{array}{c} \vdots \\ \circlearrowleft j_2 \\ \circlearrowleft i_1 \quad \circlearrowright j_1 \\ \circlearrowright i_2 \\ \vdots \\ a_x \end{array}, \quad \{i_1, i_2, j_1, j_2\} \in \{0, 1\}. \quad (4.1)$$

The four indices placed around the vertex represent its *edge states*. We refer to i_1, i_2 as *incoming* states, whereas j_1, j_2 are called *outgoing*. These indices take values in $\{0, 1\} = \{\text{empty}, \text{occupied}\}$. Whenever an edge state is equal to 1 we draw an up-oriented or right-oriented *path* on that edge, but leave the edge empty if its state is equal to 0. A vertex centered at position (x, y) also has two parameters a_x, a_y associated to it and is assigned a Boltzmann weight that depends on its edge states and

which is a function of the product $a_x a_y$. Of the 16 possible edge state configurations, only six receive nonzero Boltzmann weights: these are precisely the six that exhibit conservation of paths passing through the vertex. We list their weights below:

1	1	$\frac{1 - a_x a_y}{1 - t a_x a_y}$	$\frac{t(1 - a_x a_y)}{1 - t a_x a_y}$	$\frac{(1 - t)a_x a_y}{1 - t a_x a_y}$	$\frac{1 - t}{1 - t a_x a_y}$

(4.2)

The Boltzmann weights are stochastic in the following sense.

PROPOSITION 4.1 (Stochasticity)

Let $w_{a_x a_y}(i_1, i_2; j_1, j_2)$ be the weight of the vertex in (4.1). The parameters a_x, a_y, t can be chosen such that $0 \leq w_{a_x a_y}(i_1, i_2; j_1, j_2) \leq 1$ for all $\{i_1, i_2, j_1, j_2\}$, and for any fixed i_1, i_2 , we have

$$\sum_{j_1, j_2 \in \{0, 1\}} w_{a_x a_y}(i_1, i_2; j_1, j_2) \equiv 1. \tag{4.3}$$

Proof

The first property holds if we assume that a_x, a_y, t are real and satisfy $0 \leq a_x, a_y, t < 1$. The four cases $\{i_1, i_2\} = \{0, 0\}, \{1, 1\}, \{0, 1\}, \{1, 0\}$ of (4.3) can be easily checked, using the vertex weights (4.2). □

A *corner vertex* is a vertex formed by the union of a north and west edge (with the omission of south and east edges). They have the following form:

$$\begin{array}{c}
 \text{---} \\
 \text{---} \\
 \bullet \\
 \text{---} \\
 \text{---} \\
 \bullet \\
 \text{---} \\
 \text{---} \\
 \text{---}
 \end{array}
 \quad , \quad \{i, j\} \in \{0, 1\}. \tag{4.4}$$

a_x

There are two types of corner vertices to which we assign Boltzmann weight 1: those in which a path enters from the left and is absorbed at the center of the vertex, with no path emerging from the top; and those in which no path enters from the left, with a path being created at the center of the vertex and then emerging from the top. The remaining two corner vertex configurations are disallowed and receive a weight of 0:

$$\begin{array}{cccc}
 \begin{array}{c} \uparrow \\ \cdots a_x \\ a_x \end{array} & \begin{array}{c} \vdots \\ a_x \longrightarrow \\ a_x \end{array} & \begin{array}{c} \vdots \\ \cdots a_x \\ a_x \end{array} & \begin{array}{c} \uparrow \\ a_x \longrightarrow \\ a_x \end{array} \\
 1 & 1 & 0 & 0
 \end{array} \quad (4.5)$$

In contrast to bulk vertices, the Boltzmann weight assigned to a corner vertex is independent of the parameter a_x which is attached to it.

The Boltzmann weights are chosen as in (4.2) and (4.5) because they satisfy the Yang–Baxter equation (bulk vertices) and a boundary Yang–Baxter or reflection equation (corner vertices).

PROPOSITION 4.2 (Yang–Baxter equation)

For any fixed $0 \leq i_1, i_2, i_3, j_1, j_2, j_3 \leq 1$, we have

$$\begin{aligned}
 & \sum_{0 \leq k_1, k_2, k_3 \leq 1} w_{a_z/a_y}(i_1, i_2; k_1, k_2) w_{a_x a_z}(k_1, i_3; j_1, k_3) w_{a_x a_y}(k_2, k_3; j_2, j_3) \\
 &= \sum_{0 \leq k_1, k_2, k_3 \leq 1} w_{a_x a_y}(i_2, i_3; k_2, k_3) w_{a_x a_z}(i_1, k_3; k_1, j_3) w_{a_z/a_y}(k_1, k_2; j_1, j_2),
 \end{aligned}$$

where a_x, a_y, a_z are three arbitrary parameters. (w_{a_z/a_y} is obtained from $w_{a_x a_y}$ by substituting a_z/a_y in place of $a_x a_y$ in (4.2).)

Proof

This is a classical result in statistical mechanics (see [9]). It can also be checked by direct computation, although in this case there are 2^6 individual equations to verify. \square

PROPOSITION 4.3 (Reflection equation)

Let the weight of the corner vertex in (4.4) be denoted by $c(i; j) = \delta_{i,1-j}$. For any fixed $0 \leq i_1, i_2, j_1, j_2 \leq 1$, we have

$$\begin{aligned}
 & \sum_{0 \leq k_1, k_2, \ell_1, \ell_2 \leq 1} w_{a_y/a_x}(i_1, i_2; k_1, k_2) c(k_1; \ell_1) w_{a_x a_y}(k_2, \ell_1; \ell_2, j_1) c(\ell_2; j_2) \\
 &= \sum_{0 \leq k_1, k_2, \ell_1, \ell_2 \leq 1} c(i_2; k_2) w_{a_x a_y}(i_1, k_2; k_1, \ell_2) c(k_1; \ell_1) w_{a_y/a_x}(\ell_2, \ell_1; j_2, j_1),
 \end{aligned} \quad (4.6)$$

where a_x, a_y are two arbitrary parameters. (w_{a_y/a_x} is obtained from $w_{a_x a_y}$ by substituting a_y/a_x in place of $a_x a_y$ in (4.2).)

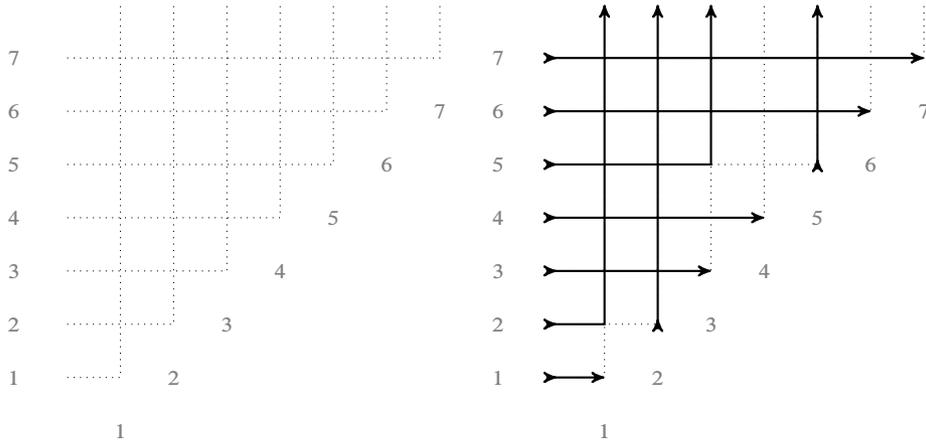


Figure 4. Left: The half-quadrant. Right: Sample configuration \mathcal{C} of the stochastic six-vertex model in the half-quadrant, for which $h(7, 7) = 4$. The seventh path string is given by $\sigma_7(\mathcal{C}) = (1, 1, 1, 0, 1, 0, 0)$.

- One can repeat this procedure to fill out the whole of the half-quadrant by induction on n .

This Markovian procedure defines the stochastic six-vertex model on the half-quadrant. Equivalently, one can think of this procedure as inducing a probability measure on random configurations of paths in the half-quadrant: the probability of the cylindric set of the path that starts off some fixed finite configuration near the origin is just the product of the Boltzmann weights of the vertices in that configuration. The latter point of view will be especially useful in what follows.

4.3. Height function and path-string distribution

The height function h is a random variable defined on the vertices of the half-quadrant. For all (x, y) such that $x \leq y$, we define

$$h(x, y) = \text{number of paths that cross one of the vertices } (i, y) \text{ for } 1 \leq i \leq x. \tag{4.7}$$

More generally, we will be interested in the distribution of up-right paths which exit to the north of the n th horizontal gridline, for some $n \geq 1$. This collection of paths forms a random binary string (s_1, \dots, s_n) , where each s_k equals 1 if the north edge of the vertex (k, n) is occupied and equals 0 otherwise. We refer to it as the n th path string. If \mathcal{C} is a configuration of the stochastic six-vertex model on the half-quadrant, then we let $\sigma_n(\mathcal{C})$ denote its n th path string (see the right panel of Figure 3). The problem of finding the distribution of $\sigma_n(\mathcal{C})$ is equivalent to calculating the partition

In view of (4.8), this theorem straight away yields the following corollary.

COROLLARY 4.5

Let $(\ell(\lambda^{(i)}))_{1 \leq i \leq n}$ be the lengths of partitions in an ascending half-space Hall–Littlewood process (4.9), and let $(h(i, n))_{1 \leq i \leq n}$ be the values of the height function along the n th horizontal line in the half-space six-vertex model. These two random vectors are equally distributed:

$$\mathbb{P}_{\text{PHL}}((\ell(\lambda^{(i)}))_{1 \leq i \leq n} = (k_i)_{1 \leq i \leq n}) = \mathbb{P}((h(i, n))_{1 \leq i \leq n} = (k_i)_{1 \leq i \leq n}).$$

The rest of the section is devoted to the proof of Theorem 4.4. It proceeds along parallel lines to a proof in [16], relating the distribution of lengths of partitions in an (ordinary) ascending Hall–Littlewood process to the distribution of the six-vertex model height function in the full quadrant $\mathbb{Z}_{\geq 0}^2$. Both the proof in [16] and the proof in the present article are extensions of ideas that were developed in [63], where integrability in a model of t -deformed bosons was used to prove refined Cauchy- and Littlewood-type summation identities involving Hall–Littlewood polynomials. Similar ideas in the context of slightly more general, higher-spin, six-vertex models were developed independently in [13] and [22].

Remark 4.6

One could prove a slightly more general version of Theorem 4.4 relating a not necessarily ascending half-space Hall–Littlewood process with the height distribution in a six-vertex model in a more complicated domain (jagged domain) following along the same lines as in the proof of [16, Theorem 5.6]. This is not useful for our present purposes, and we do not pursue it.

4.4. An integrable model of t -bosons

Following [16] and [63], we consider another integrable model of up-right paths, in which horizontal edges of the lattice can be occupied by at most one path, but no restriction is imposed on the number of paths that traverse a vertical edge. Assuming conservation of lattice paths through a vertex, four types of vertices are possible. We indicate these below, along with their associated Boltzmann weights:

m

 m
 1

$m - 1$

 m
 a

$m + 1$

 m
 $(1 - t^{m+1})$

m

 m
 a

(4.11)

$$\begin{array}{cccc}
 \begin{array}{c} m \\ \uparrow \uparrow \uparrow \\ \text{---} \\ \downarrow \downarrow \downarrow \\ m \\ b \end{array} &
 \begin{array}{c} m-1 \\ \uparrow \\ \text{---} \\ \downarrow \downarrow \downarrow \\ m \\ 1 \end{array} &
 \begin{array}{c} m+1 \\ \uparrow \uparrow \uparrow \\ \text{---} \\ \downarrow \downarrow \downarrow \\ m \\ b(1-t^{m+1}) \end{array} &
 \begin{array}{c} m \\ \uparrow \uparrow \uparrow \\ \text{---} \\ \downarrow \downarrow \downarrow \\ m \\ 1 \end{array}
 \end{array} \tag{4.13}$$

We use a dark gray background to indicate that this normalization is employed, rather than that of (4.11).⁴

4.5. Row operators and their exchange relations

For all integers $i \geq 1$, let V_i be an infinite-dimensional vector space with basis vectors $\{|m\rangle_i\}_{m \in \mathbb{Z}_{\geq 0}}$. Its dual space V_i^* is spanned by $\{\langle m|_i\}_{m \in \mathbb{Z}_{\geq 0}}$, where $\langle m|_i|n\rangle_i = \delta_{m,n}$ for all $m, n \in \mathbb{Z}_{\geq 0}$. Further, we let $V_{1\dots L}$ denote the tensor product $\otimes_{i=1}^L V_i$. Joining L of the vertices (4.11) with common spectral parameter a horizontally (likewise for the vertices (4.13)) and summing over all possible states on internal horizontal edges, we obtain a row vertex. We denote the Boltzmann weight of a row vertex as shown below:

$$\begin{array}{c}
 w_a \left(\begin{array}{c} n_L \quad \dots \quad n_1 \\ \uparrow \quad \uparrow \quad \uparrow \\ \text{---} \\ \downarrow \quad \downarrow \quad \downarrow \\ m_L \quad \dots \quad m_1 \\ i \quad \dots \quad j \end{array} \right) \\
 =: w_a(i, \{m_1, \dots, m_L\} | j, \{n_1, \dots, n_L\}),
 \end{array} \tag{4.14}$$

$$\begin{array}{c}
 w_b \left(\begin{array}{c} n_L \quad \dots \quad n_1 \\ \uparrow \quad \uparrow \quad \uparrow \\ \text{---} \\ \downarrow \quad \downarrow \quad \downarrow \\ m_L \quad \dots \quad m_1 \\ i \quad \dots \quad j \end{array} \right) \\
 =: \bar{w}_b(i, \{m_1, \dots, m_L\} | j, \{n_1, \dots, n_L\}).
 \end{array} \tag{4.15}$$

We then construct row operators that act linearly on $V_{1\dots L}$ as follows:

$$\begin{aligned}
 & T_a(i|j) : |n_1\rangle_1 \otimes \dots \otimes |n_L\rangle_L \\
 & \mapsto \sum_{m_1, \dots, m_L \geq 0} w_a(i, \{m_1, \dots, m_L\} | j, \{n_1, \dots, n_L\}) |m_1\rangle_1 \otimes \dots \otimes |m_L\rangle_L,
 \end{aligned}$$

⁴Diagrams and figures in color, in which red is used instead of dark gray, are included in the screen-enhanced version of this article at <https://doi.org/10.1215/00127094-2018-0019>.

$$= \sum_{k_1, k_2 \in \{0,1\}} \sum_{p_1, p_2, \dots \geq 0} \begin{array}{c} b \rightarrow \\ \dots \\ a \leftarrow \end{array} \begin{array}{c} \color{red}{\text{■}} \quad \color{red}{\text{■}} \quad \color{red}{\text{■}} \quad \color{red}{\text{■}} \\ \dots \\ \color{gray}{\text{■}} \quad \color{gray}{\text{■}} \quad \color{gray}{\text{■}} \quad \color{gray}{\text{■}} \end{array} \begin{array}{c} n_2 \quad n_1 \\ p_2 \quad p_1 \\ k_1 \quad j_2 \\ k_2 \quad j_1 \\ m_2 \quad m_1 \end{array} \quad (4.18)$$

where bosonic vertices on the same row have the same spectral parameter (as indicated on the left of each row). The four possibilities in (4.17) are given by the four possible choices of $j_1, j_2 \in \{0, 1\}$ in (4.18).

Proof

These identities are all relations in the Yang–Baxter algebra satisfied by the matrix entries $T_a(i|j)$ and $\bar{T}_a(i|j)$, in the limit $L \rightarrow \infty$, assuming the parameters a, b satisfy $|ab| < 1$. For more details of their derivation, using the same notation as in the present article, see [16]. □

4.6. A boundary relation

In this section we note another property of the t -boson model, namely, a reflection equation that it satisfies with respect to a particular choice of boundary (see [63]).

PROPOSITION 4.9

Let $n \geq 0$ be any nonnegative integer, and fix $i, j \in \{0, 1\}$. The following identity holds:

$$\sum_{m=0}^{\infty} \prod_{k=1}^m (1 - t^{2k-1}) \times w_a \left(\begin{array}{c} n \\ i \bullet \text{---} \color{gray}{\text{■}} \text{---} j \\ 2m \end{array} \right) = \sum_{m=0}^{\infty} \prod_{k=1}^m (1 - t^{2k-1}) \times w_a \left(\begin{array}{c} n \\ i \text{---} \color{red}{\text{■}} \text{---} \bullet j \\ 2m \end{array} \right), \quad (4.19)$$

where the dot has a path-annihilating/path-creating property: if a rightward-pointing path approaches the dot from its left, no path will emerge to the right of the dot. (Similarly, if no path approaches the dot from its left, a rightward-pointing path will emerge to the right of the dot.)

Proof

We need to check the four possible values for the pair (i, j) . In each case, the value of $2m$ at the base of the vertex is completely determined by path conservation, so the infinite sums over m trivialize. Below we list the four cases.

Case 1: $i = j = 0$. On the left-hand side of (4.19), the vertex vanishes unless $2m + 1 = n$. On the right-hand side, the vertex vanishes unless $2m = n + 1$. In either case, we see that n must be odd; otherwise, both sides vanish identically. When n is odd, we have

$$\begin{aligned} & \prod_{k=1}^{(n-1)/2} (1 - t^{2k-1}) \times w_a \left(\begin{array}{c} n \\ \text{---} \bullet \text{---} \uparrow \\ \text{---} \bullet \text{---} \text{---} \\ n-1 \end{array} \right) \\ &= \prod_{k=1}^{(n+1)/2} (1 - t^{2k-1}) \times w_a \left(\begin{array}{c} n \\ \text{---} \bullet \text{---} \uparrow \\ \text{---} \bullet \text{---} \text{---} \\ n+1 \end{array} \right), \end{aligned}$$

with the equality obviously holding thanks to the vertex weights (4.11) and (4.13).

Case 2: $i = 0, j = 1$. By the conservation of paths, the left- and right-hand sides of (4.19) vanish unless $2m = n$. The equation then reads

$$\begin{aligned} & \prod_{k=1}^{n/2} (1 - t^{2k-1}) \times w_a \left(\begin{array}{c} n \\ \text{---} \bullet \text{---} \uparrow \\ \text{---} \bullet \text{---} \rightarrow \\ n \end{array} \right) \\ &= \prod_{k=1}^{n/2} (1 - t^{2k-1}) \times w_a \left(\begin{array}{c} n \\ \text{---} \bullet \text{---} \uparrow \\ \text{---} \bullet \text{---} \rightarrow \\ n \end{array} \right), \end{aligned}$$

and the equality of the two sides is immediate, since both vertices have weight equal to a .

Case 3: $i = 1, j = 0$. Similarly to Case 2, one finds that the left- and right-hand sides of (4.19) vanish unless $2m = n$. This yields the identity

$$\begin{aligned} & \prod_{k=1}^{n/2} (1 - t^{2k-1}) \times w_a \left(\begin{array}{c} n \\ \uparrow \\ \bullet \\ \uparrow \\ n \end{array} \right) \\ &= \prod_{k=1}^{n/2} (1 - t^{2k-1}) \times w_a \left(\begin{array}{c} n \\ \uparrow \\ \bullet \\ \uparrow \\ n \end{array} \right), \end{aligned}$$

which holds because the vertices on both sides have Boltzmann weight equal to 1.

Case 4: $i = j = 1$. This case is analogous to Case 1. Again we find that both sides of (4.19) vanish unless n is odd, and in the situation where n is odd, one has

$$\begin{aligned} & \prod_{k=1}^{(n+1)/2} (1 - t^{2k-1}) \times w_a \left(\begin{array}{c} n \\ \uparrow \\ \bullet \\ \uparrow \\ n+1 \end{array} \right) \\ &= \prod_{k=1}^{(n-1)/2} (1 - t^{2k-1}) \times w_a \left(\begin{array}{c} n \\ \uparrow \\ \bullet \\ \uparrow \\ n-1 \end{array} \right), \end{aligned}$$

where the equality can be easily checked using the Boltzmann weights (4.11) and (4.13). □

For a partition $\lambda = 1^{m_1} 2^{m_2} \dots$, we introduce the shorthand notation

$$\langle \lambda | = \bigotimes_{i=1}^{\infty} \langle m_i |_i, \quad | \lambda \rangle = \bigotimes_{i=1}^{\infty} | m_i \rangle_i.$$

Proposition 4.9 is a local relation on t -boson vertices, which can be extended to a global relation in the following way.

PROPOSITION 4.10

Define a boundary covector

4.7. One-variable skew Hall–Littlewood polynomials

Comparing the Boltzmann weights (4.11) and (4.13) used in the row-to-row operators with the explicit form of the one-variable skew Hall–Littlewood polynomials, one obtains the following lemma.

LEMMA 4.11 ([16, Lemma 5.3])

The matrix elements of the operators $A(a)$, $B(a)$, $\bar{C}(b)$, $\bar{D}(b)$ are one-variable skew Hall–Littlewood polynomials:

$$\langle \lambda | A(a) | \mu \rangle = (\mathbb{1}_{\ell(\lambda) = \ell(\mu)}) P_{\lambda/\mu}(a), \quad (4.22)$$

$$\langle \lambda | B(a) | \mu \rangle = (\mathbb{1}_{\ell(\lambda) = \ell(\mu) + 1}) P_{\lambda/\mu}(a),$$

$$\langle \mu | \bar{C}(b) | \lambda \rangle = (\mathbb{1}_{\ell(\lambda) = \ell(\mu) + 1}) Q_{\lambda/\mu}(b), \quad (4.23)$$

$$\langle \mu | \bar{D}(b) | \lambda \rangle = (\mathbb{1}_{\ell(\lambda) = \ell(\mu)}) Q_{\lambda/\mu}(b),$$

where λ and μ are any two partitions.

Remark 4.12

In light of Lemma 4.11, the exchange relation in Proposition 4.8 corresponds to the skew Cauchy identity (2.3) in the case $q = 0$ when Hall–Littlewood polynomials are evaluated in a single variable, while the exchange relation in Proposition 4.10 corresponds to the skew Littlewood identity (2.8).

PROPOSITION 4.13

Let $S = (s_1, \dots, s_n)$ denote a binary string, with each $s_k \in \{0, 1\}$. The support distribution (defined in Theorem 4.4) in the ascending half-space Hall–Littlewood process can be written in the form

$$\begin{aligned} \mathbb{P}\text{HL}([\lambda^{(1)} \subset \dots \subset \lambda^{(n)}]) &= (s_1, \dots, s_n) \\ &= \frac{1}{\Phi(a_1, \dots, a_n)} \times \langle \text{ev} | \prod_{k=1}^n O_k^S(a_k) | \emptyset \rangle, \end{aligned} \quad (4.24)$$

where we have defined

$$O_k^S(a) = \begin{cases} A(a) & s_k = 0, \\ B(a) & s_k = 1, \end{cases}$$

and where the product in (4.24) is ordered from right to left as the index k increases.

Proof

We insert the identity $\sum_{\lambda^{(k)}} |\lambda^{(k)}\rangle \langle \lambda^{(k)}|$ at the left of each operator $O_k^S(a_k)$, for all

$1 \leq k \leq n$. By using the formulas (4.22) for the one-variable skew Hall–Littlewood polynomials, this produces the sum

$$\begin{aligned} & \langle \text{ev} | \prod_{k=1}^n O_k^S(a_k) | \emptyset \rangle \\ &= \sum_{\lambda^{(1)} \subset \dots \subset \lambda^{(n)} = \lambda} \mathbb{1}_{\lambda' \text{ even}} \cdot b_\lambda^{\text{el}} \prod_{k=1}^n (P_{\lambda^{(k)}/\lambda^{(k-1)}}(a_k) \mathbb{1}_{\ell(\lambda^{(k)}) - \ell(\lambda^{(k-1)}) = s_k}), \end{aligned}$$

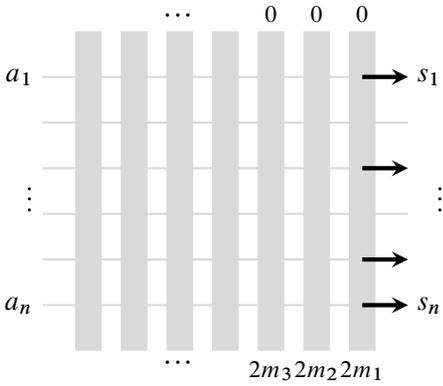
which when divided by $\Phi(a_1, \dots, a_n)$ recovers precisely the claimed distribution. \square

4.8. *Equivalence between support and path-string distributions*

We are now ready to prove Theorem 4.4, the direct equivalence between the support distribution in the ascending half-space Hall–Littlewood process and the path-string distribution in the half-space stochastic six-vertex model.

Proof of Theorem 4.4

We start by expressing the support distribution as an expectation value in the t -boson model, as in (4.24). Graphically, this relation takes the form

$$\begin{aligned} & \mathbb{P}^{\text{PHL}}([\lambda^{(1)} \subset \dots \subset \lambda^{(n)}] = (s_1, \dots, s_n)) \\ &= \prod_{1 \leq i < j \leq n} \left(\frac{1 - a_i a_j}{1 - t a_i a_j} \right) \\ & \times \sum_{m_1, m_2, \dots}^* \prod_{i=1}^\infty \prod_{k=1}^{m_i} (1 - t^{2k-1}) \end{aligned}$$


(4.25)

where the boundary conditions of the partition function are as follows.

- (1) No paths enter from the left edge of the lattice, which is considered to be infinitely far to the left.

- (2) $2m_i$ paths enter from the bottom of the i th column, where columns are counted from right to left.
- (3) $s_i \in \{0, 1\}$ paths leave via the right boundary of the i th row, where rows are counted from top to bottom.
- (4) No paths leave from the top of the lattice.

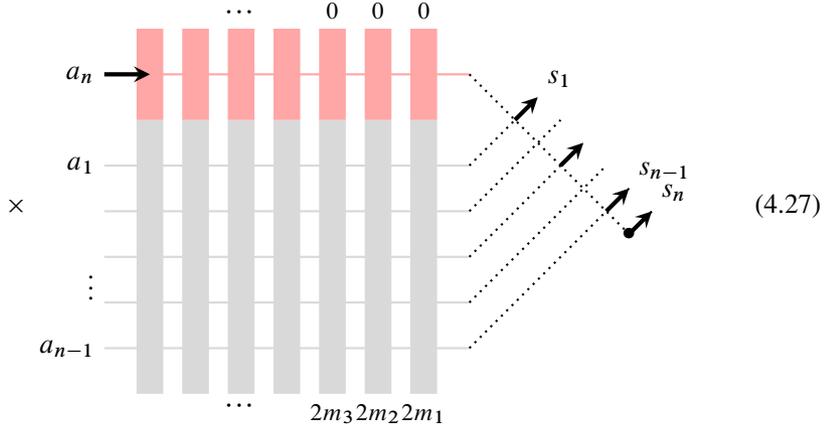
Now consider the n th row of the lattice. Depending on the value of s_n , it either encodes the operator $A(a_n)$ (in the case $s_n = 0$) or the operator $B(a_n)$ (in the case $s_n = 1$). Irrespective of the value of s_n , we will be able to apply one of the relations (4.20) to convert this operator into $\bar{D}(a_n)$ or $\bar{C}(a_n)$. This takes us to the partition function

$$\begin{aligned} & \text{PHL}([\lambda^{(1)} \subset \dots \subset \lambda^{(n)}] = (s_1, \dots, s_n)) \\ &= \prod_{1 \leq i < j \leq n} \left(\frac{1 - a_i a_j}{1 - t a_i a_j} \right) \\ & \times \sum_{m_1, m_2, \dots}^* \prod_{i=1}^{\infty} \prod_{k=1}^{m_i} (1 - t^{2k-1}) \end{aligned} \tag{4.26}$$

The diagram shows a lattice with columns and rows. The bottom row is labeled a_n and has arrows pointing to s_n . The top row is labeled a_1 and has arrows pointing to s_1 . The columns are labeled with m_i at the bottom and 0 at the top. The lattice is shaded with dark gray and light gray.

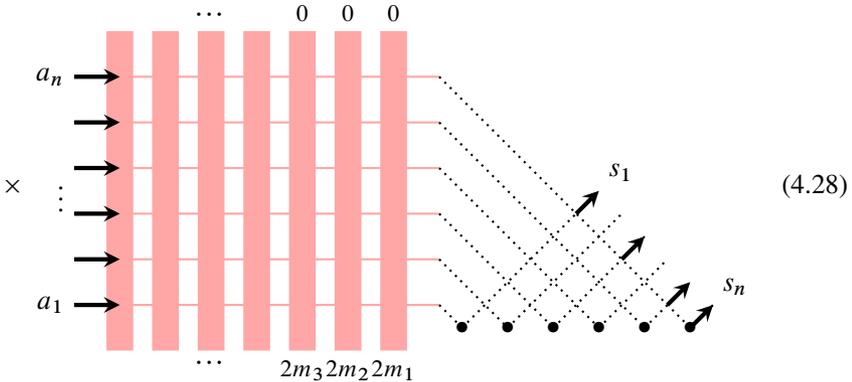
Noting that the factor $\prod_{i=1}^{n-1} (1 - a_i a_n) / (1 - t a_i a_n)$ is present in (4.26), we can now use the relation (4.18) n times to transfer the dark gray row to the top of the lattice. The result of this procedure is

$$\begin{aligned} & \text{PHL}([\lambda^{(1)} \subset \dots \subset \lambda^{(n)}] = (s_1, \dots, s_n)) \\ &= \prod_{1 \leq i < j \leq n-1} \left(\frac{1 - a_i a_j}{1 - t a_i a_j} \right) \\ & \times \sum_{m_1, m_2, \dots}^* \prod_{i=1}^{\infty} \prod_{k=1}^{m_i} (1 - t^{2k-1}) \end{aligned}$$



We can then iterate the steps in (4.26) and (4.27) a further $n - 1$ times. This converts all light gray rows into dark gray ones:

$$\begin{aligned} \text{PHL}([\lambda^{(1)} \subset \dots \subset \lambda^{(n)}]) &= (s_1, \dots, s_n) \\ &= \sum_{m_1, m_2, \dots}^* \prod_{i=1}^{\infty} \prod_{k=1}^{m_i} (1 - t^{2k-1}) \end{aligned}$$



Now we observe that the bosonic lattice is completely trivialized, since the n incoming paths at the left edge cannot leave the lattice via the top external edges. This forces them to propagate horizontally, so that the dark gray horizontal lines are completely saturated by paths. This in turn means that each m_i must be equal to 0, trivializing the summation. The resulting frozen lattice configuration has weight 1, so we remove it and retain only the six-vertex partition function emerging from the right:

$$\mathbb{P}^{\text{PHL}}([\lambda^{(1)} \subset \dots \subset \lambda^{(n)}] = \{s_1, \dots, s_n\})$$

$$= \tag{4.29}$$

The partition function appearing on the right-hand side evaluates the probability that the n th path string in the stochastic six-vertex model is equal to $\{s_1, \dots, s_n\}$, that is, $\mathbb{P}(\sigma_n(\mathcal{C}) = \{s_1, \dots, s_n\})$, completing the proof of (4.10). \square

COROLLARY 4.14

For any $x \in \mathbb{R}$, $n \in 2\mathbb{Z}_{>0}$, and any parameters $(a_1, \dots, a_n) \in (0, 1)^n$,

$$\mathbb{E}\left[\frac{1}{(-t^{x+n-h(n,n)}, t^2)_\infty}\right] = \text{Pf}[\mathbf{J} + \mathbf{f}_x \cdot \mathbf{K}^{\mathcal{G}}]_{\ell^2(\mathbb{Z}_{\geq 0})}, \tag{4.30}$$

where $h(n, n)$ is the height function at (n, n) in the half-space stochastic six-vertex model and \mathbf{f}_x is defined in (3.7).

Proof

This follows immediately from Proposition 3.3 and Corollary 4.5. \square

Remark 4.15

Corollary 4.14 could be used to show an analogue of Theorem A for the height function of the half-space stochastic six-vertex model, that is, proving that $h(n, n)$ has Tracy–Widom GOE fluctuations as n goes to infinity. A possible approach would be to adapt the arguments of [3, Appendix B] to the half-space case, but we do not pursue that here.

5. Half-line ASEP with open boundary

Consider the six-vertex model in a half-quadrant from Section 4.1 where $a_x \equiv a$, and scale a as

$$a = 1 - \frac{(1-t)\varepsilon}{2}, \quad \varepsilon \xrightarrow{\varepsilon > 0} 0,$$

so that to first order in ε ,

$$\begin{aligned} \mathbb{P}\left(\begin{array}{c} \vdots \\ \text{---} \\ \vdots \end{array}\right) &\approx \varepsilon, & \mathbb{P}\left(\begin{array}{c} \vdots \\ \text{---} \\ \vdots \\ \text{---} \\ \vdots \end{array}\right) &\approx 1 - \varepsilon, \\ \mathbb{P}\left(\begin{array}{c} \vdots \\ \text{---} \\ \vdots \\ \text{---} \\ \vdots \end{array}\right) &\approx t\varepsilon, & \mathbb{P}\left(\begin{array}{c} \vdots \\ \text{---} \\ \vdots \\ \text{---} \\ \vdots \\ \text{---} \\ \vdots \end{array}\right) &\approx 1 - t\varepsilon. \end{aligned} \tag{5.1}$$

Moreover, we rescale n as $n = \tau\varepsilon^{-1}$ with finite $\tau > 0$.

Definition 5.1

The *half-line ASEP* is a continuous-time Markov process on the state space parameterized by occupation variables $\{(\eta_x)_{x \in \mathbb{Z}_{>0}} \in \{0, 1\}^{\mathbb{Z}_{>0}}\}$. The state $\eta(\tau)$ at time τ evolves according to the following dynamics: at any given time $\tau \in [0, \infty)$ and $x \in \mathbb{Z}_{>0}$, a particle jumps from site x to $x + 1$ at exponential rate

$$\eta_x(\tau)(1 - \eta_{x+1}(\tau)) \in \{0, 1\}$$

and jumps from site $x + 1$ to x at exponential rate

$$t \eta_{x+1}(\tau)(1 - \eta_x(\tau)) \in \{0, t\}.$$

Further, a particle is created or annihilated at site 1 at exponential rates

$$\alpha(1 - \eta_1(\tau)) \quad \text{and} \quad \gamma \eta_1(\tau).$$

All these events are independent. We will restrict our attention to the case $\alpha = 1/2$ and $\gamma = t/2$ (see Figure 5) and to the empty initial condition where there are no particles in the system at time 0. We define the current at site x by

$$N_x(\tau) = \sum_{i=x}^{\infty} \eta_i(\tau),$$

and we simply denote by $N(\tau) = N_1(\tau)$ the number of particles in the system at time τ .

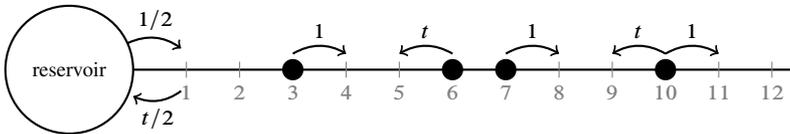


Figure 5. Jump rates in the half-line ASEP.

PROPOSITION 5.2

Under the scalings and boundary and initial conditions as in Definition 5.1 (with $\alpha = 1/2$ and $\gamma = t/2$), for any $x \in \{1, 2, \dots\}$,

$$n - x - \mathfrak{h}(n - x, n) \xrightarrow[\varepsilon \rightarrow 0]{} N_x(\tau),$$

where \mathfrak{h} is defined in (4.7) and $N_x(\tau)$ is defined in Definition 5.1. Moreover, along a sequence of ε 's such that n is even,

$$n - \mathfrak{h}(n, n) \xrightarrow[\varepsilon \rightarrow 0]{} \lceil N(\tau) \rceil_2,$$

where for an integer k , we define

$$\lceil k \rceil_2 := \min\{i \in 2\mathbb{Z} : i \geq k\}.$$

Remark 5.3

For the stochastic six-vertex model in a quadrant and the ASEP on \mathbb{Z} , a heuristic approach to the convergence was provided in [19]. Consider the ensemble of paths in the six-vertex model, and interpret each path as the trajectory of a particle where the vertical axis is the time. Under the scalings that we consider, the dynamics of this particle system converge to those of the ASEP (modulo a shift of particle positions by time). The convergence was proven rigorously in [1] for general initial conditions. The main difficulty is that, for a system with infinitely many particles, the distribution of one particle may depend on the position of other particles far away. This is why [1] considers versions of both models restricted to an interval $[-N, M]$ (which consequently have finitely many particles), proves that the convergence holds for the bounded models, and shows that the unbounded models are well approximated by the bounded ones as M and N go to infinity. In our case, the only additional complication is the boundary, which we address in the below proof.

Proof of Proposition 5.2

For the empty initial condition that we consider in this article, we will show that the tail distribution of the total number of particles injected by the reservoir into the system decays exponentially fast, uniformly in ε , so that the convergence of the dynamics of finitely many particles is enough to prove Proposition 5.2. Hence, we will focus on carefully justifying that the boundary behavior in the six-vertex model converges to that of the half-line ASEP as in Definition 5.1.

The first step is to apply a particle-hole inversion to the ensemble of paths defined by the stochastic six-vertex model in a half-quadrant (see Figure 6). This means that edges of the lattice occupied by a path become empty and empty edges become occupied by a path. Next we associate to the (particle-hole transformed) ensemble of paths

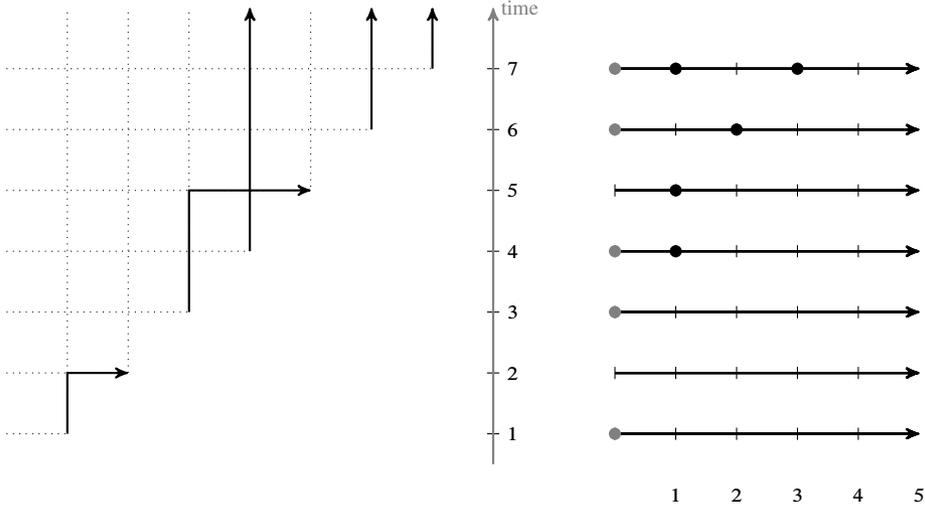


Figure 6. Left: The same arrow configuration as in Figure 2 after particle-hole transformation.

Right: Corresponding particle configurations. The presence of the particle in gray at 0 is completely determined by the rest of the configuration, ensuring that the total number of particles and the time have the same parity. A particle may be injected at site 1 in the next step only when there is a gray particle at 0. A particle may be removed from the system in the next step only when there is no gray particle at 0.

the evolution of a particle configuration. Let $\zeta_i(s)$ (viewed as an occupation variable at position i and time s) be 1 if there is an outgoing vertical arrow out of vertex $(s - i, s)$ and 0 otherwise (see Figure 6). Denote by

$$y_1 > y_2 > \cdots > y_{M(s)} > 0$$

the associated positions of the $M(s)$ particles occupying positive sites. We do not record the presence of the particle at 0, because it can be deduced from the parity of $s + M(s)$. (Indeed, $s + M(s) + \zeta_0(s)$ is always even.) Note that $M(s) = s - 1 - \mathfrak{h}(s - 1, s)$, and for even s , one can write $\lceil M(s) \rceil_2 = s - \mathfrak{h}(s, s)$.

The particle configuration $(y_i)_{i \in \mathbb{Z}_{>0}}$ can be constructed as a discrete-time Markov process where particles are updated from right to left—that is, y_1 , then y_2 , and so on—according to the following rules.

- (1) For $1 \leq i \leq M(s)$, the i th particle jumps by 1 at time $s + 1$ with probability $\varepsilon + o(\varepsilon)$, provided that $y_{i-1}(s + 1) > y_i(s) + 1$ (with the convention that $y_0 = \infty$).
- (2) For $1 \leq i \leq M(s)$ and for all $1 \leq j \leq y_i(s) - y_{i+1}(s) - 1$ such that $y_i(s) - j > 0$, the i th particle jumps by $-j$ at time $s + 1$ with probability $(t\varepsilon)^j + o(\varepsilon^j)$.

- (3) If $M(s) + s$ is odd and $y_{M(s)}(s) > 1$, then a new particle is created at $y_{M(s+1)}(s+1) = 1$ and $M(s+1) = M(s) + 1$, happening with probability $\varepsilon + o(\varepsilon)$.
- (4) If $M(s) + s$ is odd and $y_{M(s)}(s) = k > 0$, then this particle ejects from the system at time $s+1$ and $M(s+1) = M(s) - 1$ with probability $(t\varepsilon)^k + o(\varepsilon^k)$.
- (5) Each particle stays put with the complementary probability.

Let $x_i^\varepsilon(\tau) := y_i(\tau\varepsilon^{-1})$ be the position of the particles at large times. Let $I_{6v}(s)$ (resp., $I(\tau)$) be the total number of particles injected in the six-vertex process ζ (resp., the ASEP process η) between times 0 and s (resp., 0 and τ). Observe that the tail distribution of $I_{6v}(s)$ can be naively bounded by

$$\mathbb{P}(I_{6v}(\tau\varepsilon^{-1}) > x) < \mathbb{P}(\mathfrak{Bin}(\varepsilon, \tau\varepsilon^{-1}) > x) < e^{-c\varepsilon}, \quad (5.2)$$

for some constant $c > 0$ independent of ε , where $\mathfrak{Bin}(p, n)$ denotes a binomial random variable with parameters p and n . If, for any $x, k \in \mathbb{Z}_{\geq 0}$,

$$\mathbb{P}(M(\tau\varepsilon^{-1}) = x \ \& \ I_{6v}(\tau\varepsilon^{-1}) = k) \xrightarrow{\varepsilon \rightarrow 0} \mathbb{P}(N(\tau) = x \ \& \ I(\tau) = k), \quad (5.3)$$

then we may sum over k and use (5.2) to conclude that

$$\mathbb{P}(M(\tau\varepsilon^{-1}) = x) \xrightarrow{\varepsilon \rightarrow 0} \mathbb{P}(N(\tau) = x).$$

Now we turn to the proof of (5.3). Fix $k \in \mathbb{Z}_{\geq 0}$, and let us restrict our analysis to the event where $I_{6v}(\tau\varepsilon^{-1}) = k$, so that we need to consider the dynamics of only finitely many particles $x_1^\varepsilon(\tau), \dots, x_k^\varepsilon(\tau)$. It is clear (from the convergence of the discrete-time random walk on \mathbb{Z}^d to a continuous-time one or, more precisely, from [1]) that dynamics of particles away from the boundary converge to the ASEP. According to rule (3) of the dynamics, whenever site 1 is empty, a particle is added to the system at site 1 after a random time that converges to an exponential with rate 1/2. (This 1/2 comes from the fact that the particle creation is not possible at all times but only half of them because of the parity condition.) Similarly, according to rule (4) of the dynamics, whenever site 1 is occupied, this particle is removed from the system after a random time that converges to an exponential with rate $t/2$. Hence, we have shown that, on the event where $I_{6v}(\tau\varepsilon^{-1}) = k$, the joint distribution of $x_1^\varepsilon(\tau), \dots, x_k^\varepsilon(\tau)$ converges as ε goes to 0 to the distribution of ASEP particles $x_1(\tau), \dots, x_k(\tau)$, which implies (5.3). One can similarly deduce the convergence of $n - x - \mathfrak{h}(n - x, n)$ to $N_x(\tau)$ for any $x \in \{1, 2, \dots\}$. \square

PROPOSITION 5.4

For any time $\tau > 0$ and $x \in \mathbb{R}$,

$$\mathbb{E} \left[\frac{1}{(-t^x + \lceil N(\tau) \rceil_2, t^2)_\infty} \right] = \text{Pf}[\mathbf{J} + \mathbf{f}_x \cdot \mathbf{K}^{\text{ASEP}}]_{\ell^2(\mathbb{Z}_{\geq 0})}, \quad (5.4)$$

where f_x is defined in (3.7) and

$$\begin{aligned} \mathbf{K}_{11}^{\text{ASEP}}(u, v) &= \frac{1}{(2i\pi)^2} \iint \frac{(w-z)g(z)g(w)}{(z^2-1)(w^2-1)(zw-1)} \frac{dz}{z^u} \frac{dw}{w^v} \\ &\quad + \frac{1}{2i\pi} \int \frac{g(z)}{z^2-1} \frac{dz}{z^v} \mathbb{1}_{u \in 2\mathbb{Z}} \\ &\quad - \frac{1}{2i\pi} \int \frac{g(z)}{z^2-1} \frac{dz}{z^u} \mathbb{1}_{v \in 2\mathbb{Z}} + r(u, v), \end{aligned} \quad (5.5a)$$

$$\begin{aligned} \mathbf{K}_{12}^{\text{ASEP}}(u, v) &= \frac{1}{(2i\pi)^2} \iint \frac{(w-z)g(z)g(w)}{(z^2-1)w(zw-1)} \frac{dz}{z^u} \frac{dw}{w^v} \\ &\quad + \frac{1}{2i\pi} \int \frac{g(z)}{z^{v+1}} dz \mathbb{1}_{u \in 2\mathbb{Z}}, \end{aligned} \quad (5.5b)$$

$$\mathbf{K}_{22}^{\text{ASEP}}(u, v) = \frac{1}{(2i\pi)^2} \iint \frac{(w-z)g(z)g(w)}{zw(zw-1)} \frac{dz}{z^u} \frac{dw}{w^v}, \quad (5.5c)$$

where $g(z) = \exp\left(\frac{(1-t)\tau}{2} \frac{z+1}{z-1}\right)$, $r(u, v)$ is defined in (3.5), and the contours are chosen as positively oriented circles with radius less than 1.

Proof

Recall that, from Corollary 4.14,

$$\mathbb{E}^{\text{PHL}} \left[\frac{1}{(-t^x + n - \mathfrak{h}(n, n), t^2)_\infty} \right] = \text{Pf}[\mathbf{J} + f_x \cdot \mathbf{K}^{\mathfrak{G}}]_{\ell^2(\mathbb{Z}_{\geq 0})}. \quad (5.6)$$

Fix $\tau > 0$, and consider an even integer n and the corresponding ε such that $n = \tau\varepsilon^{-1}$. Recall that a is a parameter also depending on n through ε . We will let n go to infinity along the even integers. Under the scalings considered, with $f(z) = \left(\frac{z-a}{1-az}\right)^n$ as in Lemma 3.2,

$$(-1)^n f(z) \xrightarrow{n \rightarrow \infty} g(z)$$

uniformly on compact subsets of $\mathbb{C} \setminus \{1\}$, and we can discard this factor $(-1)^n$, since n is assumed even.

By Proposition 5.2, the left-hand side of (5.6) converges to the left-hand side of (5.4) as n goes to infinity. (Note that the observables under expectations have values in $(0, 1)$; thus, pointwise convergence is sufficient.)

On the other hand, $\mathbf{K}^{\mathfrak{G}}$ converges pointwise to \mathbf{K}^{ASEP} as $\varepsilon \rightarrow 0$. To conclude that the Fredholm Pfaffians also converge, it is enough to check that the Fredholm determinant expansion is absolutely convergent. The integrand of the kernel is uniformly (in ε) bounded on its contour and the integrals are on finite contours, so the kernel

is bounded. One concludes using Hadamard's bound (see Lemma 3.1) to control the Fredholm Pfaffian expansion. \square

Remark 5.5

We could define a Pfaffian point process with correlation kernel \mathbf{K}^{ASEP} . The Pfaffian point process describes the edge of the Pfaffian Schur measure under the specific limit that we considered. For the (determinantal) Schur measure, a similar limit was computed in [21] and called the *discrete Laguerre ensemble* due to its close connection to Laguerre orthogonal polynomials.

6. Fluctuations of the current in half-line ASEP

Definition 6.1

The *GOE Tracy–Widom distribution* (see [56]) is a continuous probability distribution on \mathbb{R} with cumulative distribution function given by (see, e.g., [5, Section 2.3])

$$F_{\text{GOE}}(x) = \text{Pf}[\mathbf{J} - \mathbf{K}^{\text{GOE}}]_{\mathbb{L}^2(x, \infty)},$$

where \mathbf{K}^{GOE} is the 2×2 matrix-valued kernel defined by

$$\mathbf{K}_{11}^{\text{GOE}}(x, y) = \frac{1}{(2i\pi)^2} \int dz \int dw \frac{z-w}{z+w} e^{z^3/3+w^3/3-xz-yw}, \quad (6.1a)$$

$$\begin{aligned} \mathbf{K}_{12}^{\text{GOE}}(x, y) &= -\mathbf{K}_{21}^{\text{GOE}}(x, y) \\ &= \frac{1}{(2i\pi)^2} \int dz \int dw \frac{w-z}{2w(z+w)} e^{z^3/3+w^3/3-xz-yw} \\ &\quad + \frac{1}{2} \frac{1}{2i\pi} \int dz e^{z^3/3-xz}, \end{aligned} \quad (6.1b)$$

$$\begin{aligned} \mathbf{K}_{22}^{\text{GOE}}(x, y) &= \frac{1}{(2i\pi)^2} \int dz \int dw \frac{z-w}{4zw(z+w)} e^{z^3/3+w^3/3-xz-yw} \\ &\quad + \frac{1}{2i\pi} \int e^{z^3/3-zx} \frac{dz}{4z} - \frac{1}{2i\pi} \int e^{z^3/3-zy} \frac{dz}{4z} - \frac{\text{sgn}(x-y)}{4}, \end{aligned} \quad (6.1c)$$

where all integration contours are $\mathcal{C}_1^{\pi/3}$, the contour formed by the union of two semi-infinite rays departing 1 with angles $\pi/3$ and $-\pi/3$, oriented from $1 + \infty e^{-i\pi/3}$ to $1 + \infty e^{+i\pi/3}$. Moreover, there exists a Pfaffian point process on \mathbb{R} having this kernel, and we call this point process the *GOE point process*. We will denote its points by $\mathfrak{a}_1 > \mathfrak{a}_2 > \dots$.

THEOREM 6.2

For any $t \in [0, 1)$,

$$\lim_{T \rightarrow \infty} \mathbb{P} \left(\frac{N\left(\frac{T}{1-t}\right) - \frac{T}{4}}{2^{-4/3} T^{1/3}} > -x \right) = F^{\text{GOE}}(x). \quad (6.2)$$

Proof

First we remark that $\lceil N(\tau) \rceil_2$ and $N(\tau)$ differ by at most one and consequently have the same limit under the scalings considered in the statement of the theorem. Recall that, from Proposition 5.4,

$$\mathbb{E} \left[\frac{1}{(-t^y + \lceil N(\tau) \rceil_2, t^2)_\infty} \right] = \text{Pf}[\mathbf{J} + \mathbf{f}_y \cdot \mathbf{K}^{\text{ASEP}}]_{\ell^2(\mathbb{Z}_{\geq 0})}. \quad (6.3)$$

Define the scaling function

$$\mathfrak{s}(x) := \frac{T}{4} - 2^{-4/3} x T^{1/3}.$$

We can let $\tau = \frac{T}{1-t}$ and scale y as $y = -\mathfrak{s}(\kappa)$ for a fixed $\kappa \in \mathbb{R}$. Assume that under this scaling the right-hand side of (6.3) converges for every $\kappa \in \mathbb{R}$ to the distribution function $F^{\text{GOE}}(\kappa)$. Then, by using [17, Lemma 4.1.39], the random variable

$$\frac{N\left(\frac{T}{1-t}\right) - \frac{T}{4}}{2^{-4/3} \kappa T^{1/3}} \quad (6.4)$$

weakly converges to a random variable with distribution function $F^{\text{GOE}}(\kappa)$ as T goes to infinity. We now consider the right-hand side of (6.3); the function \mathbf{f}_y multiplying the kernel is asymptotically an indicator function.

LEMMA 6.3

For a fixed $\kappa \in \mathbb{R}$, the function $\mathbb{R} \rightarrow (-1, 0)$ defined by

$$\mathbf{f}_\kappa^{(T)} : x \mapsto \mathbf{f}_{\mathfrak{s}(\kappa)}(\mathfrak{s}(x))$$

converges pointwise to $x \mapsto -\mathbb{1}_{x > \kappa}$.

Proof

We have

$$\mathbf{f}_\kappa^{(T)}(x) = \mathbf{f}_{\mathfrak{s}(\kappa)}(\mathfrak{s}(x)) = \frac{(-t^{1+2^{-4/3}T^{1/3}(\kappa-x)}; t^2)_\infty}{(-t^{2^{-4/3}T^{1/3}(\kappa-x)}; t^2)_\infty} - 1.$$

As T goes to infinity, if $\kappa > x$, then the first argument in the q -Pochhammer symbols goes to 0, so that the q -Pochhammer symbols themselves converge to 1 and \mathbf{f} goes to 0. If $\kappa < x$, we need to show that the ratio of q -Pochhammer symbols converges to 0. Let

$$p(x) := \frac{(-t^{1+x}; t^2)_\infty}{(-t^x; t^2)_\infty}.$$

We have

$$p(x)p(x+1) = \frac{1}{1+t^x},$$

which, since $p(x)$ is increasing, shows that $p(x)$ goes to 0 as x goes to $-\infty$. Hence, $\mathbf{f}_{\mathfrak{s}(\kappa)}(\mathfrak{s}(x))$ goes to -1 as T goes to infinity when $\kappa < x$. \square

Thus, our main task is to compute the limit of \mathbf{K}^{ASEP} . We will use Laplace's method to find the asymptotics of the kernel in (5.5a), (5.5b), and (5.5c). The limit of (6.3) as T goes to infinity can be computed as the limit of

$$\text{Pf}[\mathbf{J} + \mathbf{f}_\kappa^{(T)} \cdot \mathbf{K}^{(T)}]_{\ell^2(\mathbb{D}_T)},$$

where $\mathbf{K}^{(T)}$ is the rescaled kernel

$$\begin{aligned} & \mathbf{K}^{(T)}(x, y) \\ & := \frac{(-1)^{\mathfrak{s}(x)+\mathfrak{s}(y)}}{2^{-4/3}T^{1/3}} \\ & \quad \times \begin{pmatrix} \mathbf{K}_{11}^{\text{ASEP}}(\mathfrak{s}(x), \mathfrak{s}(y)) & 2^{-4/3}T^{1/3}\mathbf{K}_{12}^{\text{ASEP}}(\mathfrak{s}(x), \mathfrak{s}(y)) \\ 2^{-4/3}T^{1/3}\mathbf{K}_{21}^{\text{ASEP}}(\mathfrak{s}(x), \mathfrak{s}(y)) & (2^{-4/3}T^{1/3})^2\mathbf{K}_{22}^{\text{ASEP}}(\mathfrak{s}(x), \mathfrak{s}(y)) \end{pmatrix}, \end{aligned} \quad (6.5)$$

and the domain \mathbb{D}_T of the Fredholm Pfaffian is defined so that $\mathfrak{s}(\mathbb{D}_T) = \mathbb{Z}_{\geq 0}$. The presence of the factor $(-1)^{\mathfrak{s}(x)+\mathfrak{s}(y)}$ is a technical convenience whose purpose shall be explained later. It does not change the value of the Fredholm Pfaffian, because this extra factor has the same effect as a conjugation of the kernel. (Since $\mathfrak{s}(x), \mathfrak{s}(y) \in \mathbb{Z}$, $(-1)^{\mathfrak{s}(x)+\mathfrak{s}(y)} = (-1)^{\mathfrak{s}(x)-\mathfrak{s}(y)} = (-1)^{-\mathfrak{s}(x)+\mathfrak{s}(y)} = (-1)^{-\mathfrak{s}(x)-\mathfrak{s}(y)}$.) The power of $2^{-4/3}T^{1/3}$ that multiplies each entry of \mathbf{K}^{ASEP} is determined so that each entry remains bounded as T goes to infinity, and the global factor $(2^{-4/3}T^{1/3})^{-1}$ in front of the kernel will disappear when we approximate discrete sums by integrals.

Let us first formally examine the limit of $\mathbf{K}^{(T)}$. In the formulas (5.5) for \mathbf{K}^{ASEP} , the parameter T appears in the function $g(z)$ and in the variables u, v through the scalings that we consider. If $\mathfrak{s}(x) = \frac{T}{4} - 2^{-4/3}T^{1/3}x$ as above, we can rewrite the factors depending on T using

$$\frac{g(z)}{z^{\mathfrak{s}(x)}} = \exp(TG(z) + 2^{-4/3}T^{1/3}x \log(z)), \quad (6.6)$$

where

$$G(z) = \frac{1}{2} \frac{z+1}{z-1} - \frac{1}{4} \log(z),$$

and the branch cut of the logarithm is taken on the positive reals. The function $z \mapsto \Re[G(z)]$ will control the asymptotics of the kernel. One can check that $G'(-1) = G''(-1) = 0$ and $G'''(-1) = 1/8$, so that the Taylor expansion of G at -1 is

$$TG(-1 + \tilde{z}T^{-1/3}) \approx -T \frac{\mathbf{i}\pi}{4} + \frac{2^{-4}(\tilde{z})^3}{3}.$$

Moreover,

$$2^{-4/3}T^{1/3}x \ln(-1 + \tilde{z}T^{-1/3}) = 2^{-4/3}T^{1/3}x\mathbf{i}\pi - 2^{-4/3}x\tilde{z} + \mathcal{O}(T^{-1/3}),$$

so that for $z = 1 + \tilde{z}T^{-1/3}$,

$$\frac{g(z)}{z^{\mathfrak{s}(x)}} = \exp\left(\mathfrak{s}(x)\mathbf{i}\pi + \frac{2^{-4}(\tilde{z})^3}{3} - 2^{-4/3}x\tilde{z}\right) + \mathcal{O}(T^{-1/3}).$$

The factor $\exp(\mathfrak{s}(x)\mathbf{i}\pi)$ will be canceled by the $(-1)^{\mathfrak{s}(x)}$ in the definition of $\mathbf{K}^{(T)}$ in (6.5), and this is why we have multiplied the kernel by a factor $(-1)^{\mathfrak{s}(x)+\mathfrak{s}(y)}$. Thus, under the scalings considered, in the neighborhood of the critical point, the first integrand in the formula for $\mathbf{K}_{11}^{\text{ASEP}}$ will converge pointwise to

$$\frac{\tilde{z} - \tilde{w}}{4\tilde{z}\tilde{w}(\tilde{z} + \tilde{w})} \exp\left(\frac{\sigma^3\tilde{z}^3}{3} + \frac{\sigma^3\tilde{w}^3}{3} - \sigma\tilde{z}x - \sigma\tilde{w}y\right),$$

where $\sigma = 2^{-4/3}$ and we have rescaled the integration variables as $z = -1 + \tilde{z}T^{-1/3}$ and likewise for w . It is easy to see that all other integrands converge pointwise to similar formulas as well, and we will produce the formulas later. However, the factors $\mathbb{1}_{u \in 2\mathbb{Z}}$, $\mathbb{1}_{v \in 2\mathbb{Z}}$, and $\mathbb{1}_{v-u \in 2\mathbb{Z}+1}$ that appear in (5.5a) and (5.5b) do not have a pointwise limit as T goes to infinity, since they depend on the parity of $\mathfrak{s}(x)$ and $\mathfrak{s}(y)$.

Since we are interested in the convergence of $\text{Pf}[\mathbf{J} - \mathbf{f}_k^{(T)} \cdot \mathbf{K}^{(T)}]$, we do not need to prove the pointwise convergence of $\mathbf{K}^{(T)}$. Instead, we will first search for a kernel \mathbf{K}^∞ defined on $\mathbb{L}^2(\kappa, \infty)$ for any $\kappa \in \mathbb{R}$ such that, for any compact Lebesgue measurable set $\mathbb{K} \subset \mathbb{R}^k$,

$$\sum_{x_1, \dots, x_k \in \mathbb{K}} \text{Pf}(\mathbf{K}^{(T)}(x_i, x_j))_{i,j=1}^k \xrightarrow{T \rightarrow \infty} \int_{\mathbb{K}} dx_1 \cdots dx_k \text{Pf}(\mathbf{K}^\infty(x_i, x_j))_{i,j=1}^k, \quad (6.7)$$

where the sum on the left-hand side is over all k -tuples $(x_1, \dots, x_k) \in \mathbb{K}$ such that, for all $1 \leq i \leq k$, $\mathfrak{s}(x_i) = \frac{T}{4} - 2^{-4/3}T^{1/3}x_i \in \mathbb{Z}$. If $k = 1$, one may replace the indicator functions in $\mathbf{K}^{(T)}$ by their average value; that is,

$$\mathbb{1}_{u \in 2\mathbb{Z}} \approx \frac{1}{2}, \quad \mathbb{1}_{v \in 2\mathbb{Z}} \approx \frac{1}{2}.$$

One has to be careful that, in general, the average of a product is not the product of averages. However, for any $k \geq 1$, if we expand the Pfaffian on the left-hand side of

(6.7) as in (3.1), we note that indicator functions involving x_i can be multiplied by an indicator function involving x_j only when $j \neq i$, so that we can approximate all indicator functions by their average value. More precisely, we will use

$$\begin{aligned} (-1)^{\mathfrak{s}(x)} \mathbb{1}_{\mathfrak{s}(x) \in 2\mathbb{Z}} &\Rightarrow \frac{1}{2}, \\ (-1)^{\mathfrak{s}(x)+\mathfrak{s}(y)} r(\mathfrak{s}(x), \mathfrak{s}(y)) &\Rightarrow \frac{1}{4} \operatorname{sgn}(y-x), \end{aligned}$$

where \Rightarrow means that the convergence holds in the integrated sense of (6.7). Indeed, we have

$$r(u, v) = \frac{1}{4}((-1)^u - (-1)^v) + \frac{1}{2} \operatorname{sgn}(v-u) \mathbb{1}_{v-u \in 2\mathbb{Z}+1}.$$

The quantities $(-1)^u$ and $(-1)^v$ have average 0 and the quantity $(-1)^{u+v} \mathbb{1}_{v-u \in 2\mathbb{Z}+1}$ has average $-1/2$, so that $(-1)^{u+v} r(u, v)$ has average $\frac{1}{4} \operatorname{sgn}(u-v)$. Finally, $\operatorname{sgn}(\mathfrak{s}(x) - \mathfrak{s}(y)) = \operatorname{sgn}(y-x)$.

We claim that Laplace's method shows that (6.7) holds with \mathbf{K}^∞ being the kernel obtained by approximating all integrals in $\mathbf{K}^{(T)}$ for z and w in a neighborhood of size $T^{-1/3}$ around the critical point, taking the pointwise limit of integrands, and replacing indicator functions by their average values as prescribed above. In order to justify this rigorously, we need to check two facts.

- (1) The integration in (5.5) can be restricted to a neighborhood of -1 of arbitrarily small size, making an error going to 0 as T goes to infinity uniformly in x, y .
- (2) All integrands in $\mathbf{K}^{(T)}$ can be indeed approximated by their pointwise limit, making an error going to 0.

To prove (1), it is enough to show that the contour employed in \mathbf{K}^{ASEP} can be freely deformed to a finite contour \mathcal{C} satisfying the following property: for any $\eta > 0$, there exist $\xi > 0$ such that, for all $z \in \mathcal{C}$ with $|z+1| > \eta$, we have $\Re[G(z)] < -\xi$. This will result in the integrand being exponentially small as T goes to infinity outside of an η -neighborhood around the critical point. One readily checks that the circle of radius 1 centered at 0 is a contour line for $\Re[G(z)]$. Moreover, the Taylor expansion at -1 implies that there is also a contour line departing -1 with angles $\pm\pi/6$, and because $\Re[G(z)]$ is harmonic, it must enclose the singularity at 0 and stay inside of the circle of radius 1. Hence, one may find a contour with the desired properties between two closed contour lines (see Figure 7, where a possible choice is depicted), and Taylor expansion of G around -1 shows that this contour \mathcal{C} may depart -1 with any angles between $\pm\pi/6$ and $\pm\pi/2$. Since integrations must avoid the pole at -1 , the contour needs to be modified in a region of size $\mathcal{O}(T^{-1/3})$ around -1 , and one can choose a contour departing the point $-1 + T^{-1/3}$ with angles $\pm\pi/3$, thus leading in the limit to the contour $\mathcal{C}_1^{\pi/3}$ from Definition 6.1.

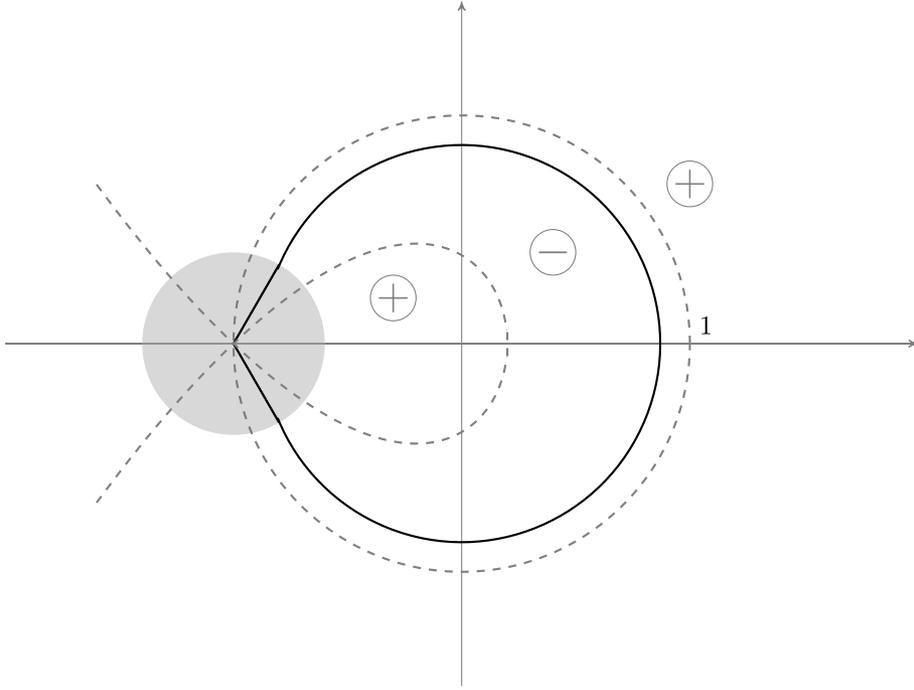


Figure 7. Contour \mathcal{C} used in the proof of Theorem 6.2 and Lemma 7.11 (in black). The dashed lines are contour lines of $\Re[G(z)]$, and we use the symbols \oplus or \ominus to indicate regions where $\Re[G(z)]$ is positive or negative. The gray area is the η -neighborhood around -1 . The part of the contour formed by two segments departing -1 is the contour used in $I_1(x)$, and the circular part is the contour used in $I_2(x)$. ($I_1(x)$ and $I_2(x)$ are defined in the proof of Lemma 7.11.)

Proving (2) amounts to controlling the error made by replacing all quantities with their Taylor approximations, and it can be done using very standard bounds (see, e.g., [5, (71) and (72)]). Since the functions G and \log admit a Taylor expansion at any order and the exponential factors of the form (6.6) are simply multiplied by rational functions in z and w , the necessary bounds are exactly the same as in previous papers (e.g., [3], [5], [18]), so we do not repeat the argument here.

At this point, we have deduced that $K^{(T)}$ converges to K^∞ in the sense of (6.7) where

$$K_{11}^\infty(x, y) = \frac{1}{(2i\pi)^2} \iint dz dw \frac{z-w}{4zw(z+w)} \exp\left(\frac{\sigma^3 \bar{z}^3}{3} + \frac{\sigma^3 w^3}{3} - \sigma z x - \sigma w y\right) - \frac{1}{2i\pi} \int \frac{dz}{4z} \exp\left(\frac{\sigma^3 z^3}{3} - \sigma z y\right)$$

$$\begin{aligned}
& + \frac{1}{2i\pi} \int \frac{dz}{4z} \exp\left(\frac{\sigma^3 z^3}{3} - \sigma z x\right) + \frac{\operatorname{sgn}(y-x)}{4}, \\
\mathbf{K}_{12}^\infty(x, y) & = \frac{1}{(2i\pi)^2} \iint dz dw \frac{z-w}{2z(z+w)} \exp\left(\frac{\sigma^3 z^3}{3} + \frac{\sigma^3 w^3}{3} - \sigma z x - \sigma w y\right) \\
& + \frac{1}{2i\pi} \int \frac{1}{2} \exp\left(\frac{\sigma^3 z^3}{3} - \sigma z y\right), \\
\mathbf{K}_{22}^\infty(x, y) & = \frac{1}{(2i\pi)^2} \iint dz dw \frac{z-w}{z+w} \exp\left(\frac{\sigma^3 z^3}{3} + \frac{\sigma^3 w^3}{3} - \sigma z x - \sigma w y\right),
\end{aligned}$$

where the contour for z and w is $\mathcal{C}_1^{\pi/3}$ (see Definition 6.1) in all integrals. In order to conclude from (6.7) that

$$\operatorname{Pf}[\mathbf{J} - \mathbf{f}_\kappa^{(T)} \cdot \mathbf{K}^{(T)}]_{\ell^2(\mathbb{D}_T)} \xrightarrow{T \rightarrow \infty} \operatorname{Pf}[\mathbf{J} - \mathbf{K}^\infty]_{\mathbb{L}^2(\kappa, \infty)},$$

one needs to estimate the entries of $\mathbf{K}^{(T)}$ and use Hadamard's bound (see Lemma 3.1) to conclude that the Fredholm Pfaffian expansion converges to the desired limit using the dominated convergence theorem. The following bounds are sufficient.

LEMMA 6.4

There exist positive constants C , c , T_0 such that, for $T > T_0$ and $x, y > \kappa$,

$$\begin{aligned}
|T^{1/3} \mathbf{K}_{11}^{(T)}(x, y)| & < C, \\
|T^{1/3} \mathbf{K}_{12}^{(T)}(x, y)| & < C \exp(-cy), \\
|T^{1/3} \mathbf{K}_{22}^{(T)}(x, y)| & < C \exp(-cx - cy).
\end{aligned}$$

Proof

These bounds are obtained in a very similar way to those in [5, Lemmas 5.11 and 6.4], where the structure of the kernel considered is the same as ours. Let us explain how the idea works for $\mathbf{K}_{22}^{(T)}$. Using the definition of the rescaled kernel $\mathbf{K}^{(T)}$ in (6.5) and the formula for $\mathbf{K}_{22}^{\text{ASEP}}$ in (5.5c), we have for some constant C ,

$$\begin{aligned}
& |T^{1/3} \mathbf{K}_{22}^{(T)}(x, y)| \\
& \leq C T^{2/3} \left| \frac{1}{(2i\pi)^2} \iint \frac{w-z}{zw(zw-1)} \right. \\
& \quad \left. \times \exp\left(T(G(z) - G(w)) + 2^{-4/3} T^{1/3} (x \log(z) - y \log(w))\right) dz dw \right|.
\end{aligned}$$

The integrations can be restricted to a neighborhood of size η around -1 , where η can be arbitrarily small, by doing an error bounded by $e^{-c_1(\eta)T - c_2(\eta)T^{1/3}x}$ for some

constants $c_1(\eta)$, $c_2(\eta)$. Now we make the change of variables $z = -1 + \tilde{z}T^{-1/3}$ and likewise for w . By using the Taylor expansion of G and approximations of the other factors in the integrand, there exists a constant C such that, for T large enough,

$$\begin{aligned} & |T^{1/3}\mathbf{K}_{22}^{(T)}(x, y)| \\ & \leq \frac{C}{(2i\pi)^2} \iint \left| \frac{z-w}{z+w} \right| \\ & \quad \times \left| \exp\left(\frac{\sigma^3 z^3}{3} + \frac{\sigma^3 w^3}{3} - \sigma z x - \sigma w y + C T^{-1/3}(z^4 + w^4)\right) \right| dz dw. \end{aligned}$$

Let $\mathcal{C}_a^{\pi/3}$ be the contour formed by two semi-infinite rays departing a with directions $\pm\pi/3$. The integration contour in the integral above can be chosen as the intersection of $\mathcal{C}_a^{\pi/3}$ —for some a that can be freely chosen in $\mathbb{R}_{>0}$ —with the ball of radius $T^{1/3}\eta$ around -1 . Note that the prefactor $T^{2/3}$ canceled with the Jacobian of the change of variables. On this contour, $T^{-1/3}z^4$ is bounded by $\eta|z|^3$, so that $\exp(T^{-1/3}z^4)$ will not compensate for the decay of $\exp(\Re[z^3])$ for η small enough. Moreover, the factor $(z-w)/(z+w)$ stays bounded along the contour. Hence, there exists a constant C such that

$$|T^{1/3}\mathbf{K}_{22}^{(T)}(x, y)| \leq \frac{C}{(2i\pi)^2} \iint |\exp(z^3/3 + w^3/3 - zx - wy)| dz dw. \quad (6.8)$$

For x and y positive, it is easy to conclude from (6.8) that $|T^{1/3}\mathbf{K}_{22}^{(T)}(x, y)|$ has exponential decay in x and y . \square

To conclude the proof of Theorem 6.2, we have to show that

$$\text{Pf}[\mathbf{J} - \mathbf{K}^\infty]_{\mathbb{L}^2(\kappa, \infty)} = \text{Pf}[\mathbf{J} - \mathbf{K}^{\text{GOE}}]_{\mathbb{L}^2(\kappa, \infty)}.$$

Note that the Fredholm Pfaffian of \mathbf{K}^∞ does not depend on σ , so that one may take $\sigma = 1$. In this case,

$$\begin{aligned} \mathbf{K}_{11}^\infty(x, y) &= \mathbf{K}_{22}^{\text{GOE}}(x, y), \\ \mathbf{K}_{12}^\infty(x, y) &= \mathbf{K}_{12}^{\text{GOE}}(y, x) = -\mathbf{K}_{21}^{\text{GOE}}(x, y), \\ \mathbf{K}_{22}^\infty(x, y) &= \mathbf{K}_{11}^{\text{GOE}}(x, y). \end{aligned}$$

Exchanging the even rows and columns with the odd rows and columns, one multiplies the Pfaffian of $(\mathbf{K}^\infty(x_i, x_k))_{i,j=1}^k$ by $(-1)^k$. By multiplying even rows and even columns by -1 , the Pfaffian gets multiplied by $(-1)^k$ another time, and one arrives at \mathbf{K}^{GOE} . \square

7. KPZ equation on \mathbb{R}_+

In [31], the half-space open ASEP with specially tuned weak asymmetry and weak boundary conditions is shown to converge to the half-space KPZ equation with Neumann-type boundary conditions. That paper deals with a set of initial data which they call “near equilibrium” and a boundary condition that makes the boundary repulsive, in a sense that we shall explain shortly. Neither the initial data (“narrow-wedge” type) nor the boundary condition with which we work is covered by the results of [31]. In fact, the type of initial data we consider requires a slightly different scaling (a logarithmic correction at the level of the height function; see Remark 7.4).

We show in this section that the observable of the ASEP (Definition 5.1) that is expected (in light of the results from [31]) to converge to the solution of the half-space KPZ equation has a weak limit whose distribution can be characterized. When the first version of this article was posted, the identification of this limit with the KPZ equation solution had not yet been proved. However, subsequently, the work of [31] was extended by [46] to include more general initial data (including narrow wedge) and a general boundary condition parameter $A \in \mathbb{R}$ (which includes the relevant case for us). We have not modified this section to reflect the work of [46] (besides adding a few parenthetical notes or footnotes). Combining our Corollary 7.7 with [46, Theorem 1.2] yields [46, Corollary 1.3], which is the one-point Laplace transform for the half-space KPZ equation from narrow-wedge initial data with boundary parameter $A = -1/2$.

Definition 7.1 (Half-space stochastic heat equation)

We say that $\mathcal{Z}(T, X)$ is a *mild solution* to the stochastic heat equation (SHE)

$$\partial_T \mathcal{Z} = \frac{1}{2} \Delta \mathcal{Z} + \mathcal{Z} \dot{W}, \quad (7.1)$$

on \mathbb{R}_+ with delta initial data at the origin and Robin boundary condition with parameter $A \in \mathbb{R}$,

$$\partial_X \mathcal{Z}(T, X)|_{X=0} = A \mathcal{Z}(T, 0) \quad (\forall T > 0), \quad (7.2)$$

if $\mathcal{Z}(T, \cdot)$ is adapted to the filtration $\sigma\{\mathcal{Z}(0, \cdot), W|_{[0, T]}\}$ and

$$\mathcal{Z}(T, X) = \mathcal{P}_T^R(X, 0) + \int_0^T \int_0^\infty \mathcal{P}_{T-S}^R(X, Y) \mathcal{Z}(S, Y) dW_S(dY), \quad (7.3)$$

where the last integral is the Itô integral with respect to the cylindrical Wiener process W and \mathcal{P}^R is the heat kernel satisfying the Robin boundary condition

$$\partial_X \mathcal{P}_T^R(X, Y)|_{X=0} = A \mathcal{P}_T^R(0, Y) \quad (\forall T > 0, Y > 0). \quad (7.4)$$

The Hopf–Cole solution to the half-space KPZ equation with Neumann boundary condition with parameter A is defined to be $\log \mathcal{Z}$. It was shown in [31] that for $A \geq 0$ the half-space SHE admits a unique solution. It has the property that almost surely $\mathcal{Z}(T, X) > 0$ for all $T > 0$ and $X \geq 0$, so that the logarithm of $\mathcal{Z}(T, X)$ is well defined. When $A > 0$, the heat kernel \mathcal{P}_T^R corresponds to the transition kernel of a Brownian motion killed at the origin at rate A times the local time. This is why the boundary condition with $A > 0$ can be classified as a repulsive boundary condition. When $A < 0$, the kernel \mathcal{P}_T^R corresponds to the transition kernel of Brownian motion duplicating at the origin at rate $|A|$ times the local time. When $A = 0$, \mathcal{P}_T^R is the transition kernel of a Brownian motion reflected at the origin.

7.1. Convergence of ASEP to the SHE

In order to relate the ASEP with the SHE, we must perform a microscopic version of the Cole–Hopf transform (also called the Gärtner transform). We will work presently with the notation of [31] and then match it to that of the present article.

Definition 7.2 (Microscopic Cole–Hopf/Gärtner transform)

Consider the ASEP with left jump rate \mathfrak{q} , right jump rate \mathfrak{p} , input rate from the reservoir α , and output rate to the reservoir γ (see Figure 1). Define a height function $h_\tau(x)$ at time τ and position $x \in \mathbb{Z}_{\geq 0}$ by

$$h_\tau(x) = h_\tau(0) + \sum_{y=1}^x \hat{\eta}_\tau(y),$$

where $h_\tau(0)$ is 2 times the net number of particles that are removed (i.e., the number of particles that move into the source minus the number that move out of the source) from site $x = 1$ during the time interval $[0, \tau]$ (in particular, $h_0(0) = 0$), and $\hat{\eta}_\tau(x) = 2\eta_\tau(x) - 1$ is $+1$ if there is a particle at x at time τ and -1 if there is not. For empty initial data $h_0(x) = -x$. For $\tau \geq 0$ and $x \in \mathbb{Z}_{\geq 0}$ define the microscopic Cole–Hopf transform of the ASEP as

$$Z_\tau(x) = \exp(-\lambda h_\tau(x) + \nu\tau),$$

where

$$\lambda = \frac{1}{2} \log \frac{\mathfrak{q}}{\mathfrak{p}}, \quad \nu = \mathfrak{p} + \mathfrak{q} - 2\sqrt{\mathfrak{p}\mathfrak{q}}.$$

The reason for the definition of $Z_\tau(x)$ is that inside $\mathbb{Z}_{\geq 0}$ (i.e., away from the boundary) this transformation of the ASEP satisfies a discrete SHE. By assuming further that the boundary rates satisfy (1.1), it is possible to extend the discrete SHE

to the boundary of $\mathbb{Z}_{\geq 0}$ in terms of a discrete Robin boundary condition (see [31, Section 3]). In particular,

$$Z_\tau(x) = \sum_{y \in \mathbb{Z}_{\geq 0}} p_\tau^R(x, y) Z_0(y) + \int_0^\tau \sum_{y \in \mathbb{Z}_{\geq 0}} p_{\tau-s}^R(x, y) dM_s(y),$$

where, for each y , the $M_s(y)$'s are explicit martingales, and $p_\tau^R(x, y)$ is the half-line discrete heat kernel with Robin boundary condition $p_\tau^R(-1, y) = \mu p_\tau^R(0, y)$ (see [31, Lemma 4.5] for an explicit formula for this). The parameter μ is related to α via the equality $\alpha = p^{3/2}(p^{1/2} - \mu\sqrt{q})(p - q)^{-1}$, where we also assume (1.1); that is, $\alpha/p + \gamma/q = 1$.

We now introduce weakly asymmetric scaling of the half-line ASEP.

Definition 7.3

Introduce a small parameter $\epsilon > 0$, and then scale

$$\begin{aligned} p &= \frac{1}{2}e^{\sqrt{\epsilon}}, & q &= \frac{1}{2}e^{-\sqrt{\epsilon}}, \\ \alpha &= \frac{p^{3/2}(\sqrt{p} - (1 - A\epsilon)\sqrt{q})}{p - q}, & \gamma &= \frac{q^{3/2}(\sqrt{q} - (1 - A\epsilon)\sqrt{p})}{q - p}. \end{aligned}$$

Write $Z_\tau^\epsilon(x)$ to denote $Z_\tau(x)$ with parameters given in terms of the above ϵ -parameterizations, and write the space-time-rescaled version of Z as

$$\mathcal{Z}^\epsilon(\hat{t}, \hat{x}) = \epsilon^{-1/2} Z_{\epsilon^{-2}\hat{t}}^\epsilon(\epsilon^{-1}\hat{x}). \quad (7.5)$$

For small ϵ , we have the approximations

$$\alpha = \frac{1}{4} + \left(\frac{3}{8} + \frac{1}{4}A\right)\sqrt{\epsilon} + \mathcal{O}(\epsilon), \quad \gamma = \frac{1}{4} - \left(\frac{3}{8} + \frac{1}{4}A\right)\sqrt{\epsilon} + \mathcal{O}(\epsilon),$$

and

$$\lambda = -\sqrt{\epsilon}, \quad \nu = \frac{\epsilon}{2} + \frac{\epsilon^2}{24} + \mathcal{O}(\epsilon^3).$$

For the ASEP with empty initial data (i.e., $h_0(x) = -x$ for $x \geq 0$), as $\epsilon \rightarrow 0$, we expect that $\mathcal{Z}^\epsilon \Rightarrow \mathcal{Z}$ as a space-time process, where \mathcal{Z} is the unique mild solution to the SHE on $\mathbb{R}_{\geq 0}$ with delta initial data and Robin boundary condition with parameter A at the origin. An analogous result was proved when $A \geq 0$ and for near-equilibrium initial data as [31, Theorem 2.17].

Remark 7.4

The scaling considered in (7.5) differs from [31, Definition 2.16] by an extra factor

$\epsilon^{-1/2}$ multiplying Z . This comes from the fact that the initial condition that we consider is not “near equilibrium,” and this extra factor ensures that $Z_{\tau=0}^\epsilon(\cdot)$ converges to the delta function. This is a very similar situation to those in the works of [10] and [4] regarding the full-space ASEP and KPZ equation. The first case requires a similar “near-equilibrium” type of initial data, while the second extends it to include step initial data. The basic argument of [4, Section 3] shows that for short times (i.e., in the scaling in which the ASEP converges to the KPZ equation) the step initial data height function becomes “near equilibrium.” Then, applying the existing convergence results of [10] one gets a consistent family of measures which can be extended back to time 0 and be shown to coincide with the desired “narrow-wedge” initial data KPZ equation. We leave a rigorous proof of this for future work (and hence we will not use the convergence result in the present article).

7.2. Matching notation

We now translate the convergence stated in Section 7.1 in terms of the notation used throughout the present article. Let us consider the ASEP considered in Definition 5.1, that is, the half-line ASEP with rates $\mathfrak{p} = 1$, $\mathfrak{q} = t$, $\alpha = 1/2$, $\gamma = t/2$. Let us set $t = e^{-2\sqrt{\epsilon}}$, and rescale the time by $2e^{-\sqrt{\epsilon}}$, so that effectively the jump rates become

$$\mathfrak{p} = \frac{1}{2}e^{\sqrt{\epsilon}}, \quad \mathfrak{q} = \frac{1}{2}e^{-\sqrt{\epsilon}}, \quad \alpha = \frac{1}{4}e^{\sqrt{\epsilon}}, \quad \gamma = \frac{1}{4}e^{-\sqrt{\epsilon}}.$$

If we match those rates with the rates from Definition 7.3, this corresponds to choosing the boundary parameter $A = -1/2 + \mathcal{O}(\epsilon)$. This is why we cannot apply the results of [31], since A is assumed to be nonnegative there. Since we have rescaled the time by $2e^{-\sqrt{\epsilon}}$, we have

$$h_\tau(x) = -2N_{x+1}(\tau e^{\sqrt{\epsilon}}/2) - x,$$

in the sense that the space-time processes have the same distribution, where $N_x(\tau)$ is defined in Definition 5.1 and $h_\tau(x)$ is defined in Definition 7.2. Then, under the scalings of Definition 7.3,

$$\begin{aligned} & \mathcal{Z}^\epsilon(\hat{\tau}, \hat{x}) \\ &= \epsilon^{-1/2} \exp(-2\sqrt{\epsilon}N_{1+\epsilon^{-1}\hat{x}}(\epsilon^{-2}\hat{\tau}e^{\sqrt{\epsilon}}/2) - \epsilon^{-1/2}\hat{x} + \epsilon^{-1}\hat{\tau}/2 + \hat{\tau}/24 + \mathcal{O}(\epsilon)). \end{aligned}$$

At this point, it is more convenient to reparameterize using $\tau = \hat{\tau}e^{\sqrt{\epsilon}}$ and $\varepsilon = 2\sqrt{\epsilon}$ so that $t = e^{-\varepsilon}$.

Definition 7.5

For $t = e^{-\varepsilon}$, under the notation of Definition 5.1, we define the space-time process

$$\mathcal{Z}_\tau^\varepsilon(x) := 2\varepsilon^{-1} \exp(-\varepsilon N_{1+4x\varepsilon^{-2}}(8\varepsilon^{-4}\tau) - 2\varepsilon^{-1}x + 2\varepsilon^{-2}\tau - \tau\varepsilon^{-1} + 7\tau/24).$$

We expect that $\mathcal{Z}^\varepsilon \Rightarrow \mathcal{Z}$ as a space-time process, where \mathcal{Z} is the unique mild solution to the SHE on $\mathbb{R}_{\geq 0}$ with delta initial data and Robin boundary condition with parameter $A = -1/2$ at the origin.

7.3. Multiplicative functional of the GOE and KPZ equation on \mathbb{R}_+

One can readily adapt Propositions 3.3 and 5.4, changing $-t^x$ into ζt^x , so that, for any $y \in \mathbb{R}$ and $\zeta \in \mathbb{C} \setminus \mathbb{R}_{>0}$,

$$\mathbb{E}\left[\frac{1}{(\zeta t^{y+\lceil N(\tau) \rceil 2}, t^2)_\infty}\right] = \text{Pf}[\mathbf{J} + \mathbf{h}_y \cdot \mathbf{K}^{\text{ASEP}}]_{\ell^2(\mathbb{Z}_{\geq 0})}, \quad (7.6)$$

where

$$\mathbf{h}_y(j) = \frac{(\zeta t^{y+j+1}; t^2)_\infty}{(\zeta t^{y+j}; t^2)_\infty} - 1. \quad (7.7)$$

We consider now the asymptotic behavior of this identity in the weakly asymmetric regime.

THEOREM 7.6

Under the scalings

$$t = e^{-\varepsilon}, \quad \tau = \frac{\varepsilon^{-3}\hat{\tau}}{1-t},$$

the random variable

$$\mathcal{U}_\varepsilon(\hat{\tau}) = \frac{t^{(N(\tau) - \frac{\varepsilon^{-3}\hat{\tau}}{4})}}{1-t^2} \quad (7.8)$$

weakly converges to a random variable $\mathcal{U}(\hat{\tau})$ such that, for $\zeta < 0$,

$$\mathbb{E}[\exp(\zeta \mathcal{U}(\hat{\tau}))] = \text{Pf}[\mathbf{J} + \mathbf{g} \cdot \mathbf{K}^{\text{GOE}}]_{\mathbb{L}^2(\mathbb{R})}, \quad (7.9)$$

where

$$\mathbf{g}(x) = \frac{1}{\sqrt{1 - \zeta e^{\hat{\sigma}x}}} - 1 \quad \text{and} \quad \hat{\sigma} = 2^{-4/3}(\hat{\tau})^{1/3}.$$

Before proving the theorem, let us further interpret the result. Note that we have

$$\begin{aligned} \bar{\mathcal{U}}_\varepsilon(T) &:= \mathcal{U}_\varepsilon((1-t)\varepsilon^{-1}T) \\ &= \frac{\varepsilon^{-1}}{2} \exp(-\varepsilon N(\varepsilon^{-4}T) + \frac{T}{4}\varepsilon^{-2} - \frac{T}{8}\varepsilon^{-1} + \frac{T}{24} + \mathcal{O}(\varepsilon)). \end{aligned}$$

On the one hand, $\bar{\mathcal{U}}_\varepsilon(\tau)$ converges⁵ to $\mathcal{U}(\tau)$, a certain random variable having a complicated yet explicit distribution. On the other hand, one may check that

$$\mathcal{Z}_\tau^\varepsilon(0) = 4 \bar{\mathcal{U}}_\varepsilon(8\tau) \exp(-\tau/24 + \mathcal{O}(\varepsilon)).$$

Thus, if we define

$$\mathcal{H}(\tau) = \log(4 \mathcal{U}(8\tau)) - \frac{\tau}{24},$$

we expect⁶ that $\mathcal{H}(\tau)$ will have the same distribution as $\mathcal{H}(\tau, 0)$ from Definition 7.1 with parameter $A = -1/2$. Furthermore, the Laplace transform of $e^{\mathcal{H}(\tau)}$ can be expressed as a multiplicative functional of the GOE process.

COROLLARY 7.7

For any $\tau > 0$ and $z > 0$,

$$\mathbb{E}\left[\exp\left(\frac{-z}{4} \exp\left(\frac{\tau}{24} + \mathcal{H}(\tau)\right)\right)\right] = \mathbb{E}\left[\prod_{i=1}^{+\infty} \frac{1}{\sqrt{1 + z \exp((\tau/2)^{1/3} \mathfrak{a}_i)}}\right],$$

where $\{\mathfrak{a}_i\}_{i=1}^\infty$ forms the GOE point process (see Definition 6.1).

This corollary is proved in Section 7.5.

7.4. Proof of Theorem 7.6

Our proof proceeds via the following three steps.

- (1) If $\mathcal{U}_\varepsilon(\hat{\tau})$ weakly converges to $\mathcal{U}(\hat{\tau})$, then the left-hand side of (7.6) converges to $\mathbb{E}[\exp(\zeta \mathcal{U}(\hat{\tau}))]$.
- (2) The right-hand side of (7.6) converges to $\text{Pf}[\mathbf{J} + \mathbf{g} \cdot \mathbf{K}^{\text{GOE}}]_{\mathbb{L}^2(\mathbb{R})}$.
- (3) The sequence $\mathcal{U}_\varepsilon(\hat{\tau})$ indeed weakly converges to $\mathcal{U}(\hat{\tau})$ whose distribution is determined by (7.9).

Step 1. Let us examine the scaling limit of the left-hand side in (7.6). We need a lemma about the asymptotics of q -Pochhammer symbols.

LEMMA 7.8

If two positive real numbers ξ and θ are related by

$$\theta = \frac{t^\xi}{1 - t^2},$$

⁵The proof of Theorem 7.6 is still valid when $\hat{\tau}$ is not a constant but converges to a constant as ε goes to 0. Hence, by letting $\hat{\tau} = (1 - t)\varepsilon^{-1}T$, we obtain that $\mathcal{U}_\varepsilon(T)$ converges to $\mathcal{U}(T)$.

⁶This is now proved in [46, Theorem 1.2].

then as t goes to 1,

$$\frac{1}{(\zeta t^\xi; t^2)_\infty} \rightarrow \exp(\zeta \theta)$$

uniformly for θ in a compact subset of $\mathbb{R}_{\geq 0}$ and for ζ in $\mathbb{R}_{\leq 0}$.

Proof

By using the q -binomial theorem with $q = t^2$,

$$\frac{1}{(\zeta t^\xi; t^2)_\infty} = \sum_{k=0}^{\infty} \frac{\zeta^k \theta^k (1-t^2)^k}{(t^2; t^2)_k}. \quad (7.10)$$

This expansion is absolutely convergent for any ζ in a compact set and t close enough to 1. We have

$$\frac{(1-t^2)^k}{(t^2, t^2)_k} \xrightarrow{t \rightarrow 1} \frac{1}{k!}.$$

We may use dominated convergence to conclude that the right-hand side of (7.10) converges to

$$\sum_{k=0}^{\infty} \zeta^k \frac{\xi^k}{k!} = \exp(\zeta \xi)$$

as desired. It is easy to control the error made in order to show that the convergence is uniform as θ varies in a compact subset of $\mathbb{R}_{\geq 0}$ and $\zeta \in \mathbb{R}_{\leq 0}$. \square

Let us scale y in (7.6) as $y = -\varepsilon^{-3} \hat{\tau}/4$, and apply Lemma 7.8 with $\xi_\varepsilon = y + [N(\tau)]_2$ and $\theta = \mathcal{U}_\varepsilon(\hat{\tau})$. If $\mathcal{U}_\varepsilon(\hat{\tau}) \xrightarrow{\varepsilon \rightarrow 0} \mathcal{U}(\hat{\tau})$, then

$$\frac{1}{(\zeta t^{y+[N(\tau)]_2}, t^2)_\infty} \xrightarrow{\varepsilon \rightarrow 0} \exp(\zeta \mathcal{U}(\hat{\tau})).$$

The left-hand side belongs to the interval $(0, 1)$, so that if $\mathcal{U}_\varepsilon(\hat{\tau})$ weakly converges to $\mathcal{U}(\hat{\tau})$, then

$$\mathbb{E} \left[\frac{1}{(\zeta t^{y+[N(\tau)]_2}, t^2)_\infty} \right] \xrightarrow{\varepsilon \rightarrow 0} \mathbb{E}[\exp(\zeta \mathcal{U}(\hat{\tau}))] \quad (7.11)$$

uniformly for $\zeta \in \mathbb{R}_{\geq 0}$.

Step 2. Now we examine the limit of the right-hand side in (7.6). We will rescale the kernel using $\mathfrak{s}(x) = \varepsilon^{-3} \hat{\tau}/4 - 2^{-4/3} \varepsilon^{-1} \hat{\tau}^{1/3} x$. Let us first examine the scaling limit of \mathfrak{h}_y under the scalings of Theorem 7.6.

LEMMA 7.9

For a fixed $\zeta \in \mathbb{C} \setminus \mathbb{R}_{>0}$ and with $y = -\varepsilon^{-3}\hat{\tau}/4$,

$$h_y(\mathfrak{s}(x)) \xrightarrow{\varepsilon \rightarrow 0} \frac{1}{\sqrt{1 - \zeta e^{\hat{\sigma}x}}} - 1 = g(x).$$

Proof

Let

$$p(a) = \frac{(\zeta t^{-\varepsilon^{-1}a+1}; t^2)_\infty}{(\zeta t^{-\varepsilon^{-1}a}; t^2)_\infty}.$$

We have

$$p(a)p(a + \varepsilon) = \frac{1}{1 - \zeta t^{-\varepsilon^{-1}a}} \xrightarrow{\varepsilon \rightarrow 0} \frac{1}{1 - \zeta e^a},$$

so that

$$p(a) \xrightarrow{\varepsilon \rightarrow 0} \frac{1}{\sqrt{1 - \zeta e^a}}. \quad \square$$

The convergence of \mathbf{K}^{ASEP} to \mathbf{K}^{GOE} in the integrated sense of (6.7) is already proved, since the scalings are exactly the same as those in the proof of Theorem 6.2, with $T = \varepsilon^{-3}\hat{\tau}$. The only thing left to check is that the dominated convergence theorem applies. More precisely, we need to show that, for any nonnegative integer k , under the scalings $y = -\varepsilon^{-3}\hat{\tau}/4$ and $T = \varepsilon^{-3}\hat{\tau}$, the function

$$(x_1, \dots, x_k) \mapsto \prod_{i=1}^k h_y(\mathfrak{s}(x_i)) \cdot \text{Pf}[\mathbf{K}^T(x_i, x_j)]_{i,j=1}^k \quad (7.12)$$

is absolutely summable on the domain⁷ \mathbb{D}_T of the Fredholm Pfaffian, uniformly as $\varepsilon \rightarrow 0$. Since $|h|$ is bounded by 1, using Lemma 6.4 to control $\mathbf{K}^T(x, y)$ as x, y approach $+\infty$, we know that (7.12) is uniformly absolutely summable on sets of the form $(\kappa, +\infty)^k$. The two next lemmas provide bounds for $h_y(\mathfrak{s}(x))$ and $\mathbf{K}^{(T)}(x, y)$ for x, y near $-\infty$, which prove that (7.12) is absolutely summable on the whole domain \mathbb{D}_T .

LEMMA 7.10

For $y = -\varepsilon^{-3}\hat{\tau}/4$, $\zeta < 0$, and any $x \in \mathbb{R}$,

$$|h_y(\mathfrak{s}(x))| \leq |\zeta|e^x.$$

⁷That is, the preimage of $\mathbb{Z}_{\geq 0}$ by the scaling function $\mathfrak{s}(\cdot)$.

Proof

When $\zeta < 0$, the function $h_y(\mathfrak{s}(x))$ is increasing in the variable x and stays in $(-1, 0)$, so that

$$(h_y(\mathfrak{s}(x)) + 1)^2 \geq (h_y(\mathfrak{s}(x)) + 1)(h_y(\mathfrak{s}(x) - \epsilon) + 1) = \frac{1}{1 - \zeta t^{-\epsilon^{-1}x+1}}.$$

It implies that

$$-h_y(\mathfrak{s}(x))(h_y(\mathfrak{s}(x)) + 2) \leq \frac{-\zeta t^{-\epsilon^{-1}x+1}}{1 - \zeta t^{-\epsilon^{-1}x+1}}.$$

Hence, since $\zeta < 0$ and $h \in (-1, 0)$,

$$0 \leq -h_y(\mathfrak{s}(x)) \leq -\zeta t^{-\epsilon^{-1}x+1} \leq |\zeta| e^x. \quad \square$$

LEMMA 7.11

There exist positive constants c, C, T_0 such that, for $T > T_0$,

$$\begin{aligned} |T^{1/3} \mathbf{K}_{11}^{(T)}(x, y)| &< |xy|^c, \\ |T^{1/3} \mathbf{K}_{12}^{(T)}(x, y)| &< |xy|^c, \quad \text{and} \\ |T^{1/3} \mathbf{K}_{22}^{(T)}(x, y)| &< |xy|^c. \end{aligned} \quad (7.13)$$

Proof

If these bounds hold uniformly as T goes to infinity, they should, in particular, hold for \mathbf{K}^{GOE} . Indeed, the entries in the kernel \mathbf{K}^{GOE} can be written as integrals of the Airy function and its derivative (see, for instance, the formulas in [57], which are equivalent to (6.1)). The Airy function can be defined as

$$\text{Ai}(x) = \int_{\mathcal{C}_1^{\pi/3}} e^{z^3/3 - zx} dz,$$

where the contour $\mathcal{C}_1^{\pi/3}$ is formed by the union of two semi-infinite rays departing 1 with angles $\pi/3$ and $-\pi/3$. Hence, the bounds (7.13) when $T = \infty$ may be deduced from the bounds on the Airy function and its derivative

$$|\text{Ai}(x)| \leq C |x|^{-1/4}, \quad |\text{Ai}'(x)| \leq C |x|^{1/4},$$

which are classical (see [45, Chapter 11, (1.08), (1.09)]) and can be proved by saddle-point analysis.

In order to prove the bounds (7.13) for the kernel $\mathbf{K}^{(T)}$ uniformly in T , we will likewise reduce the problem to the study of simpler functions that play roles analogous

to those of the Airy function and its derivative for the kernel \mathbf{K}^{ASEP} . The contours in the definition of \mathbf{K}^{ASEP} in (5.5) can all be chosen as circles with radius less than 1, so that by using the expansions

$$\frac{1}{1-zw} = \sum_{k=0}^{\infty} (zw)^k, \quad \frac{1}{1-z^2} = \sum_{k=0}^{\infty} z^{2k}, \quad \frac{1}{1-w^2} = \sum_{k=0}^{\infty} w^{2k},$$

all entries of the kernel $(-1)^{u+v}\mathbf{K}^{\text{ASEP}}(u, v)$ can be written using sums of the functions

$$A(u) = \frac{(-1)^u}{2i\pi} \int \frac{g(z)}{z^u} dz$$

and

$$B(u) = \frac{(-1)^u}{2i\pi} \int \frac{(z+1)g(z)}{z^u} dz,$$

where

$$g(z) = \exp\left(\frac{T}{2} \frac{z+1}{z-1}\right)$$

and the contour is a circle around the origin. More precisely, $A(u)$ plays a role analogous to that of the Airy function, while $B(u)$ plays a role analogous to that of the derivative of the Airy function. Then the kernel \mathbf{K}^{ASEP} has the same structure as \mathbf{K}^{GOE} provided that Airy functions are replaced by their discrete analogues and integrals are replaced by sums. Since we are interested in the rescaled kernel $\mathbf{K}^{(T)}$ instead of \mathbf{K}^{ASEP} , we are reduced to studying

$$\begin{aligned} A_T(x) &:= T^{1/3} A(\mathfrak{s}(x)) \\ &= T^{1/3} \frac{(-1)^{\mathfrak{s}(x)}}{2i\pi} \int \exp(TG(z) + 2^{-4/3} T^{1/3} x \log(z)) dz, \end{aligned} \quad (7.14)$$

where, as before,

$$G(z) = \frac{1}{2} \frac{z+1}{z-1} - \frac{1}{4} \log(z)$$

and $B_T(x) := T^{2/3} B(\mathfrak{s}(x))$. Both functions A_T and B_T have exponential decay at $+\infty$, for the same reasons as in Lemma 6.4. Hence, if we show polynomial bounds as x approaches $-\infty$ for A_T and B_T , then sums of the form

$$T^{-1/3} \sum_{y=x}^{\infty} A_T(y), \quad T^{-1/3} \sum_{y=x}^{\infty} B_T(y),$$

where the summations are over $y \in \mathbb{D}_T$, $y > x$, will also satisfy polynomial bounds and $\mathbf{K}^{(T)}$ will satisfy (7.13). Thus, we are left with showing that

$$|A_T(x)| \leq C|x|^{c_A}, \quad |B_T(x)| \leq C|x|^{c_B}$$

for some constants $C, c_A, c_B > 0$. The optimal exponents seem to be $c_A = -1/4$ and $c_B = 1/4$ —as for the Airy function and its derivative—but we will not need such precision. Since the arguments are very similar for $A_T(x)$ and $B_T(x)$ we will only explain how the bound is obtained for A_T .

We will distinguish two cases, whether $T > rx^6$ or $T \leq rx^6$, where r is a constant that we will optimize later to facilitate our analysis. First note that the circle of radius 1 centered at 0 is a contour line for $\Re\epsilon[G(z)]$; this will be used in both cases.

Case $T \leq rx^6$. In this case, we may integrate (7.14) on a contour of radius $1 - \eta$ for a small $\eta > 0$. On such a contour, the real part of $G(z)$ tends to 0 as η goes to 0. We can bound $A_T(x)$ by

$$|A_T(x)| \leq T^{1/3} \frac{1}{2i\pi} \int \exp(T \Re\epsilon[G(z)] + 2^{-4/3} T^{1/3} x \log(|z|)) dz.$$

Since this holds for any $\eta > 0$, one can let η go to 0 and conclude that, since the integral goes to 1 as η goes to 0,

$$|A_T(x)| \leq T^{1/3} \leq r^{1/3}|x|^2.$$

Case $T > rx^6$. Let $\eta > 0$ be a small parameter that we will choose more precisely later. In (7.14), the contour can be deformed to be the contour \mathcal{C} depicted in Figure 7, that is, a contour formed by two segments of length η departing -1 at angles $\pm\pi/3$ and an arc of circle around 0 joining the extremities of these two segments.

We separately treat the contributions of the integral in a neighborhood of size η around -1 and the contribution outside of that neighborhood. Let us write $A_T(x) = I_1(x) + I_2(x)$ according to this partition of the integration contour; that is, $I_1(x)$ is the integral in (7.14) where the integration is restricted to a neighborhood of -1 of size η , whereas $I_2(x)$ equals the integration outside of that neighborhood (see Figure 7). Outside of an η -neighborhood along the contour \mathcal{C} , we may use the bounds

$$|\exp(TG(z))| < \exp(-c_1\eta^3 T)$$

for some fixed constant c_1 (which does not depend on η), and

$$|\exp(2^{-4/3} T^{1/3} x \log(z))| < \exp(c_2 \eta T^{1/3} |x|)$$

for some fixed constant c_2 . Taking into account the prefactor $T^{1/3}$, we hence find that

$$|I_2(x)| < T^{1/3} \exp(-c_1 \eta^3 T + c_2 \eta T^{1/3} |x|).$$

Now we analyze $I_1(x)$. We make the change of variables $z = -1 + 2^{4/3} \tilde{z} T^{-1/3}$. Using Taylor approximations of the functions G and \log , we arrive at

$$I_1(x) = \frac{C_2}{2i\pi} \int_{\mathcal{C}_0^{\pi/3}} \exp\left(\frac{\tilde{z}^3}{3} - \tilde{z}x + E(\tilde{z}, T, x)\right) d\tilde{z},$$

where C_2 is a fixed constant, $E(\tilde{z}, T, x)$ is an error term, and the new contour can be chosen as $\mathcal{C}_0^{\pi/3}$, restricted to \tilde{z} such that $|\tilde{z}| < 2^{-4/3} T^{1/3} \eta$. By using Taylor approximations, this error term can be bounded by

$$|E(\tilde{z}, T, x)| < \eta |\tilde{z}|^3 + x \eta |\tilde{z}|.$$

At this point, we can adapt standard techniques used to estimate the Airy function $\text{Ai}(x)$ on the negative reals when x goes to $-\infty$. We refer, for instance, to Section 5 in [30] and more specifically after equation (5.6) therein. Let us make the change of variables $\tilde{z} = \sqrt{|x|} z$. For negative x , the integral becomes

$$I_1(x) = \frac{C_2 |x|^{1/2}}{2i\pi} \int \exp\left(|x|^{3/2} \left(\frac{z^3}{3} + z\right) + F(z, T, x)\right) dz,$$

where the integration contour has been changed according to the change of variables and the error term can be bounded by

$$|F(z, T, x)| < \eta |x|^{3/2} (|z|^3 + |z|).$$

The function $z \mapsto \frac{z^3}{3} + z$ has two critical points at $z = \pm i$, which yield contributions of the same order. One can make a new saddle-point approximation around both these saddle points and control the approximations using the same steps as in [30]. Without the error term $F(z, T, x)$, we could bound the integral $I_1(x)$ by a constant as in [30]. Taking into account the error term which grows as $|x|$ goes to infinity, we can only arrive at

$$|I_1(x)| < C_3 \exp(c_3 \eta |x|^{3/2}),$$

for some constants⁸ $C_3, c_3 > 0$.

Summing the contributions of the integral in (7.14) inside and outside of a η -neighborhood, we arrive at

$$|A_T(x)| = |I_1(x) + I_2(x)| < C_3 \exp(c_3 \eta |x|^{3/2}) + T^{1/3} \exp(-c_1 \eta^3 T + c_2 \eta T^{1/3} |x|).$$

⁸If we were bounding the function $B_T(x)$ instead of $A_T(x)$, then the constant C_3 would be replaced by $C_3 |x|^{c_4}$ for a certain exponent c_4 . Although this term grows as $|x|$ goes to infinity, the rest of the argument can be adapted as for A_T .

Recall that η is a real number that can be taken arbitrarily small, and recall that the constants c_1 , c_2 , and C_3 are independent of η . By letting $\eta = 1/|x|^{3/2}$, the first term is bounded, and the second term becomes

$$T^{1/3} \exp(-c_1 T/|x|^{9/2} + c_2 T^{1/3}/|x|^{1/2}).$$

We need to maximize this quantity over $T \in (r|x|^6, +\infty)$. Let us write $T = |x|^6 U$ so that we are interested in

$$\max_{U \in (r, +\infty)} \{|x|^2 U^{1/3} \exp(|x|^{3/2}(-c_1 U + c_2 U^{1/3}))\}.$$

One may choose r large enough, so that, for any $U > r$,

$$-c_1 U + c_2 U^{1/3} < -c_1 U/2,$$

in which case

$$\begin{aligned} & \max_{U \in (r, +\infty)} \{|x|^2 U^{1/3} \exp(|x|^{3/2}(-c_1 U + c_2 U^{1/3}))\} \\ & < |x|^2 \max_{U \in (0, +\infty)} \{U^{1/3} \exp(-|x|^{3/2} c_1 U/2)\} \\ & < C_4 \end{aligned}$$

for some constant C_4 . Finally, we have found that in any case

$$|A_T(x)| < \max\{r^{1/3}|x|^2, C_3 e^{c_3}, C_4\} < C|x|^2$$

for some constant C , which is the desired polynomial growth estimate. \square

Step 3. We employ the same argument as in Step 3 of the proof of Theorem 11.6 in [21], which we reproduce here.⁹ Lemma 7.8 shows that, under the same scalings as above,

$$\mathbb{E} \left[\frac{1}{(\zeta_{t^y + \lceil N(\hat{\tau}) \rceil_2, t^2})_\infty} \right] - \mathbb{E} [\exp(\zeta \mathcal{U}_\varepsilon(\hat{\tau}))] \quad (7.15)$$

goes to 0 as $\varepsilon \rightarrow 0$ uniformly for $\zeta \in \mathbb{R}_{\leq 0}$. By the compactness of the space of positive measures in the weak topology, the distribution of $\mathcal{U}_\varepsilon(\hat{\tau})$ must have limiting points as $\varepsilon \rightarrow 0$ in the space of positive measures on $\mathbb{R}_{\geq 0}$ of total mass at most 1. Let μ be one such limiting point. We have seen in Step 2 that

⁹Although the proof of Theorem 11.6 in [21] is only a sketch, Step 3 is rigorous. One can also prove Step 3 by adapting the arguments in [12, Example 5.5].

$$\mathbb{E} \left[\frac{1}{(\zeta t^{\gamma + \lceil N(\tau) \rceil_2, t^2})_\infty} \right] \xrightarrow{\varepsilon \rightarrow 0} \text{Pf}[\mathbf{J} + \mathbf{g} \cdot \mathbf{K}^{\text{GOE}}]_{\mathbb{L}^2(\mathbb{R})},$$

which together with (7.15) implies that

$$\int_{\mathbb{R}_{\geq 0}} \exp(\zeta y) \mu(dy) = \text{Pf}[\mathbf{J} + \mathbf{g} \cdot \mathbf{K}^{\text{GOE}}]_{\mathbb{L}^2(\mathbb{R})}. \quad (7.16)$$

Since positive measures with mass at most 1 are uniquely determined by their Laplace transform, the limiting point μ is unique. Finally, since the right-hand side of (7.16) goes to 1 as ζ goes to 0, μ is actually a probability measure. Denote by $\mathcal{U}(\hat{\tau})$ a random variable with distribution μ . Then $\mathcal{U}_\varepsilon(\hat{\tau})$ weakly converges to $\mathcal{U}(\hat{\tau})$ with, for $\zeta \leq 0$,

$$\mathbb{E}[\exp(\zeta \mathcal{U}(\hat{\tau}))] = \text{Pf}[\mathbf{J} + \mathbf{g} \cdot \mathbf{K}^{\text{GOE}}]_{\mathbb{L}^2(\mathbb{R})}.$$

7.5. Proof of Corollary 7.7

Since $1 + \mathbf{g}(x) = \frac{1}{\sqrt{1 + \zeta e^{\sigma x}}}$,

$$\text{Pf}[\mathbf{J} + \mathbf{g} \cdot \mathbf{K}^{\text{GOE}}]_{\mathbb{L}^2(\mathbb{R})} = \mathbb{E} \left[\prod_{i=1}^{+\infty} \frac{1}{\sqrt{1 + \zeta e^{\sigma a_i}}} \right], \quad (7.17)$$

provided that both sides are absolutely convergent. We know from the proof of Theorem 7.6 that the left-hand side of (7.17) is absolutely convergent. One can show that the infinite product on the right-hand side is absolutely convergent as well using the same estimates on \mathbf{K}^{GOE} .

Recall that $\sigma = 2^{-4/3} \hat{\tau}^{1/3}$. In light of (7.17), the statement of Corollary 7.7 is just a reformulation of Theorem 7.6.

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References

- [1] A. AGGARWAL, *Convergence of the stochastic six-vertex model to the ASEP: Stochastic six-vertex model and ASEP*, Math. Phys. Anal. Geom. **20** (2017), no. 3. MR 3589552. DOI 10.1007/s11040-016-9235-8. (43, 45)

- [2] ———, *Current fluctuations of the stationary ASEP and six-vertex model*, *Duke Math. J.* **167** (2018), 269–384. MR 3754630. DOI 10.1215/00127094-2017-0029. (6)
- [3] A. AGGARWAL and A. BORODIN, *Phase transitions in the ASEP and stochastic six-vertex model*, to appear in *Ann. Probab.*, preprint, arXiv:1607.08684v1 [math.PR]. (3, 6, 41, 52)
- [4] G. AMIR, I. CORWIN, and J. QUASTEL, *Probability distribution of the free energy of the continuum directed random polymer in $1 + 1$ dimensions*, *Comm. Pure Appl. Math.* **64** (2011), 466–537. MR 2796514. DOI 10.1002/cpa.20347. (8, 58)
- [5] J. BAIK, G. BARRAQUAND, I. CORWIN, and T. SUIDAN, *Pfaffian Schur processes and last passage percolation in a half-quadrant*, to appear in *Ann. Probab.*, preprint, arXiv:1606.00525v3 [math.PR]. (5, 6, 9, 17, 47, 52, 53)
- [6] J. BAIK and E. M. RAINS, *Algebraic aspects of increasing subsequences*, *Duke Math. J.* **109** (2001), 1–65. MR 1844203. DOI 10.1215/S0012-7094-01-10911-3. (2, 14)
- [7] ———, *The asymptotics of monotone subsequences of involutions*, *Duke Math. J.* **109** (2001), 205–281. MR 1845180. DOI 10.1215/S0012-7094-01-10921-6. (5, 6, 9)
- [8] G. BARRAQUAND, A. BORODIN, and I. CORWIN, *Half-space Macdonald processes*, preprint, arXiv:1802.08210v1 [math.PR]. (3)
- [9] R. J. BAXTER, *Exactly Solvable Models in Statistical Mechanics*, Academic Press, London, 1982. MR 0690578. (25)
- [10] L. BERTINI and G. GIACOMIN, *Stochastic Burgers and KPZ equations from particle systems*, *Comm. Math. Phys.* **183** (1997), 571–607. MR 1462228. DOI 10.1007/s002200050044. (58)
- [11] D. BETEA, M. WHEELER, and P. ZINN-JUSTIN, *Refined Cauchy/Littlewood identities and six-vertex model partition functions, II: Proofs and new conjectures*, *J. Algebraic Combin.* **42** (2015), 555–603. MR 3369568. DOI 10.1007/s10801-015-0592-3. (3, 15)
- [12] P. BILLINGSLEY, *Convergence of Probability Measures*, Wiley, New York, 1968. MR 0233396. (67)
- [13] A. BORODIN, *On a family of symmetric rational functions*, *Adv. Math.* **306** (2017), 973–1018. MR 3581324. DOI 10.1016/j.aim.2016.10.040. (29)
- [14] ———, *Stochastic higher spin six vertex model and Macdonald measures*, *J. Math. Phys.* **59** (2018), no. 023301. MR 3760963. DOI 10.1063/1.5000046. (9, 16)
- [15] A. BORODIN, A. BUFETOV, and I. CORWIN, *Directed random polymers via nested contour integrals*, *Ann. Physics* **368** (2016), 191–247. MR 3485045. DOI 10.1016/j.aop.2016.02.001. (9)
- [16] A. BORODIN, A. BUFETOV, and M. WHEELER, *Between the stochastic six vertex model and Hall–Littlewood processes*, to appear in *J. Combin. Theory Ser. A*, preprint, arXiv:1611.09486v1 [math.PR]. (9, 29, 33, 37)
- [17] A. BORODIN and I. CORWIN, *Macdonald processes*, *Probab. Theory Related Fields* **158** (2014), 225–400. MR 3152785. DOI 10.1007/s00440-013-0482-3. (3, 13, 48)
- [18] A. BORODIN, I. CORWIN, and P. FERRARI, *Free energy fluctuations for directed polymers in random media in $1 + 1$ dimension*, *Comm. Pure Appl. Math.* **67** (2014), 1129–1214. MR 3207195. DOI 10.1002/cpa.21520. (52)

- [19] A. BORODIN, I. CORWIN, and V. GORIN, *Stochastic six-vertex model*, *Duke Math. J.* **165** (2016), 563–624. [MR 3466163](#). [DOI 10.1215/00127094-3166843](#). (3, 43)
- [20] A. BORODIN, I. CORWIN, L. PETROV, and T. SASAMOTO, *Spectral theory for interacting particle systems solvable by coordinate Bethe ansatz*, *Comm. Math. Phys.* **339** (2015), 1167–1245. [MR 3385995](#). [DOI 10.1007/s00220-015-2424-7](#). (3)
- [21] A. BORODIN and G. OLSHANSKI, *The ASEP and determinantal point processes*, *Comm. Math. Phys.* **353** (2017), 853–903. [MR 3649488](#). [DOI 10.1007/s00220-017-2858-1](#). (3, 6, 47, 67)
- [22] A. BORODIN and L. PETROV, *Higher spin six vertex model and symmetric rational functions*, *Selecta Math. (N.S.)* **24** (2018), 751–874. [MR 3782413](#). [DOI 10.1007/s00029-016-0301-7](#). (29)
- [23] A. BORODIN and E. M. RAINS, *Eynard-Mehta theorem, Schur process, and their Pfaffian analogs*, *J. Stat. Phys.* **121** (2005), 291–317. [MR 2185331](#). [DOI 10.1007/s10955-005-7583-z](#). (2, 14, 17)
- [24] P. CALABRESE, P. LE DOUSSAL, and A. ROSSO, *Free-energy distribution of the directed polymer at high temperature*, *Europhys. Lett.* **90** (2010), no. 20002. [DOI 10.1209/0295-5075/90/20002](#). (8)
- [25] S. CORTEEL and L. K. WILLIAMS, *Tableaux combinatorics for the asymmetric exclusion process*, *Adv. in Appl. Math.* **39** (2007), 293–310. [MR 2352041](#). [DOI 10.1016/j.aam.2006.08.002](#). (5)
- [26] ———, *Staircase tableaux, the asymmetric exclusion process, and Askey–Wilson polynomials*, *Proc. Natl. Acad. Sci. USA* **107** (2010), 6726–6730. [MR 2630104](#). [DOI 10.1073/pnas.0909915107](#). (5)
- [27] ———, *Tableaux combinatorics for the asymmetric exclusion process and Askey–Wilson polynomials*, *Duke Math. J.* **159** (2011), 385–415. [MR 2831874](#). [DOI 10.1215/00127094-1433385](#). (5)
- [28] I. CORWIN, *The Kardar–Parisi–Zhang equation and universality class*, *Random Matrices Theory Appl.* **1** (2012), no. 1130001. [MR 2930377](#). [DOI 10.1142/S2010326311300014](#). (2)
- [29] I. CORWIN and E. DIMITROV, *Transversal fluctuations of the ASEP, stochastic six vertex model, and Hall–Littlewood Gibbsian line ensembles*, *Comm. Math. Phys.*, published electronically 4 May 2018. [DOI 10.1007/s00220-018-3139-3](#). (6)
- [30] I. CORWIN, J. QUASTEL, and D. REMENIK, *Continuum statistics of the Airy₂ process*, *Comm. Math. Phys.* **317** (2013), 347–362. [MR 3010187](#). [DOI 10.1007/s00220-012-1582-0](#). (66)
- [31] I. CORWIN and H. SHEN, *Open ASEP in the weakly asymmetric regime*, *Comm. Pure Appl. Math.*, published electronically 16 February 2018. [DOI 10.1002/cpa.21744](#). (2, 8, 10, 55, 56, 57, 58)
- [32] B. DERRIDA, M. R. EVANS, V. HAKIM, and V. PASQUIER, *Exact solution of a 1D asymmetric exclusion model using a matrix formulation*, *J. Phys. A* **26** (1993), 1493–1517. [MR 1219679](#). (5)
- [33] V. DOTSENKO, *Replica Bethe ansatz derivation of the Tracy–Widom distribution of the free energy fluctuations in one-dimensional directed polymers*, *J. Stat. Mech. Theory Exp.* **2010**, no. P07010. [DOI 10.1088/1742-5468/2010/07/P07010](#). (8)

- [34] H. G. DUHART, P. MÖRTERS, and J. ZIMMER, *The semi-infinite asymmetric exclusion process: Large deviations via matrix products*, Potential Anal. **48** (2018), 301–323. MR 3779090. DOI 10.1007/s11118-017-9635-9. (5)
- [35] S. GROSSKINSKY, *Phase transitions in nonequilibrium stochastic particle systems with local conservation laws*, Ph.D. dissertation, Technische Universität München, Munich, 2004. (5)
- [36] T. GUEUDRÉ and P. LE DOUSSAL, *Directed polymer near a hard wall and KPZ equation in the half-space*, Europhys. Lett. **100** (2012), no. 26006. DOI 10.1209/0295-5075/100/26006. (9)
- [37] T. HALPIN-HEALY and K. A. TAKEUCHI, *A KPZ cocktail—shaken, not stirred . . . toasting 30 years of kinetically roughened surfaces*, J. Stat. Phys. **160** (2015), 794–814. MR 3373641. DOI 10.1007/s10955-015-1282-1. (2)
- [38] M. KARDAR, G. PARISI, and Y. ZHANG, *Dynamic scaling of growing interfaces*, Phys. Rev. Lett. **56** (1986), 889–892. DOI 10.1103/PhysRevLett.56.889. (2)
- [39] N. KITANINE, K. K. KOZŁOWSKI, J. M. MAILLET, N. A. SLAVNOV, and V. TERRAS, *Algebraic Bethe ansatz approach to the asymptotic behavior of correlation functions*, J. Stat. Mech. Theory Exp. **2009**, no. P04003. MR 2504955. (4)
- [40] G. KUPERBERG, *Symmetry classes of alternating-sign matrices under one roof*, Ann. of Math. (2) **156** (2002), 835–866. MR 1954236. DOI 10.2307/3597283. (3, 6, 26)
- [41] T. M. LIGGETT, *Ergodic theorems for the asymmetric simple exclusion process*, Trans. Amer. Math. Soc. **213** (1975), 237–261. MR 0410986. DOI 10.2307/1998046. (4)
- [42] I. G. MACDONALD, *Symmetric Functions and Hall Polynomials*, 2nd ed., Oxford Math. Monogr., Clarendon, New York, 1995. MR 1354144. (10, 11, 12)
- [43] N. O’CONNELL, T. SEPPÄLÄINEN, and N. ZYGOURAS, *Geometric RSK correspondence, Whittaker functions and symmetrized random polymers*, Invent. Math. **197** (2014), 361–416. MR 3232009. DOI 10.1007/s00222-013-0485-9. (3)
- [44] A. OKOUNKOV and N. RESHETIKHIN, *Correlation function of Schur process with application to local geometry of a random 3-dimensional Young diagram*, J. Amer. Math. Soc. **16** (2003), 581–603. MR 1969205. DOI 10.1090/S0894-0347-03-00425-9. (2)
- [45] F. W. J. OLVER, *Asymptotics and Special Functions*, Academic Press, New York, 1974. MR 0435697. (63)
- [46] S. PAREKH, *The KPZ limit of ASEP with boundary*, preprint, arXiv:1711.05297v1 [math.PR]. (2, 8, 55, 60)
- [47] E. M. RAINS, *Correlation functions for symmetrized increasing subsequences*, preprint, arXiv:math/0006097v1 [math.CO]. (2, 16, 19)
- [48] ———, *Multivariate quadratic transformations and the interpolation kernel*, SIGMA Symmetry Integrability Geom. Methods Appl. **14** (2018), no. 019. MR 3772016. DOI 10.3842/SIGMA.2018.019. (3, 15)
- [49] T. SASAMOTO and T. IMAMURA, *Fluctuations of the one-dimensional polynuclear growth model in half-space*, J. Stat. Phys. **115** (2004), 749–803. MR 2054161. DOI 10.1023/B:JOSS.0000022374.73462.85. (2, 5, 6)

- [50] T. SASAMOTO and H. SPOHN, *Exact height distributions for the KPZ equation with narrow wedge initial condition*, Nuclear Phys. B **834** (2010), 523–542. [MR 2628936](#). [DOI 10.1016/j.nuclphysb.2010.03.026](#). (8)
- [51] T. SASAMOTO and L. WILLIAMS, *Combinatorics of the asymmetric exclusion process on a semi-infinite lattice*, J. Comb. **5** (2014), 419–434. [MR 3324896](#). [DOI 10.4310/JOC.2014.v5.n4.a1](#). (5)
- [52] E. K. SKLYANIN, *Boundary conditions for integrable quantum systems*, J. Phys. A. **21** (1988), 2375–2389. [MR 0953215](#). (26)
- [53] H. SPOHN, *Long range correlations for stochastic lattice gases in a nonequilibrium steady state*, J. Phys. A **16** (1983), 4275–4291. [MR 0732737](#). (2)
- [54] ———, “The Kardar-Parisi-Zhang equation: a statistical physics perspective” in *Stochastic Processes and Random Matrices*, Oxford Univ. Press, Oxford, 2017, 177–227. [MR 3728716](#). (2)
- [55] C. A. TRACY and H. WIDOM, *Level-spacing distributions and the Airy kernel*, Comm. Math. Phys. **159** (1994), 151–174. [MR 1257246](#). (2)
- [56] ———, *On orthogonal and symplectic matrix ensembles*, Comm. Math. Phys. **177** (1996), 727–754. [MR 1385083](#). (2, 47)
- [57] ———, *Matrix kernels for the Gaussian orthogonal and symplectic ensembles*, Ann. Inst. Fourier (Grenoble) **55** (2005), 2197–2207. [MR 2187952](#). (63)
- [58] ———, *Asymptotics in ASEP with step initial condition*, Comm. Math. Phys. **290** (2009), 129–154. [MR 2520510](#). [DOI 10.1007/s00220-009-0761-0](#). (3)
- [59] ———, *The asymmetric simple exclusion process with an open boundary*, J. Math. Phys. **54** (2013), no. 103301. [MR 3134598](#). [DOI 10.1063/1.4822418](#). (3, 5)
- [60] ———, *The Bose gas and asymmetric simple exclusion process on the half-line*, J. Stat. Phys. **150** (2013), 1–12. [MR 3018876](#). [DOI 10.1007/s10955-012-0686-4](#). (5)
- [61] N. V. TSILEVICH, *Quantum inverse scattering method for the q -boson model and symmetric functions*, Funct. Anal. Appl. **40** (2006), 207–217. [MR 2265685](#). [DOI 10.1007/s10688-006-0032-1](#). (30)
- [62] M. UCHIYAMA, T. SASAMOTO, and M. WADATI, *Asymmetric simple exclusion process with open boundaries and Askey–Wilson polynomials*, J. Phys. A **37** (2004), 4985–5002. [MR 2065218](#). [DOI 10.1088/0305-4470/37/18/006](#). (5)
- [63] M. WHEELER and P. ZINN-JUSTIN, *Refined Cauchy/Littlewood identities and six-vertex model partition functions, III: Deformed bosons*, Adv. Math. **299** (2016), 543–600. [MR 3519476](#). [DOI 10.1016/j.aim.2016.05.010](#). (9, 29, 33)

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